



Lecture 13: Animation

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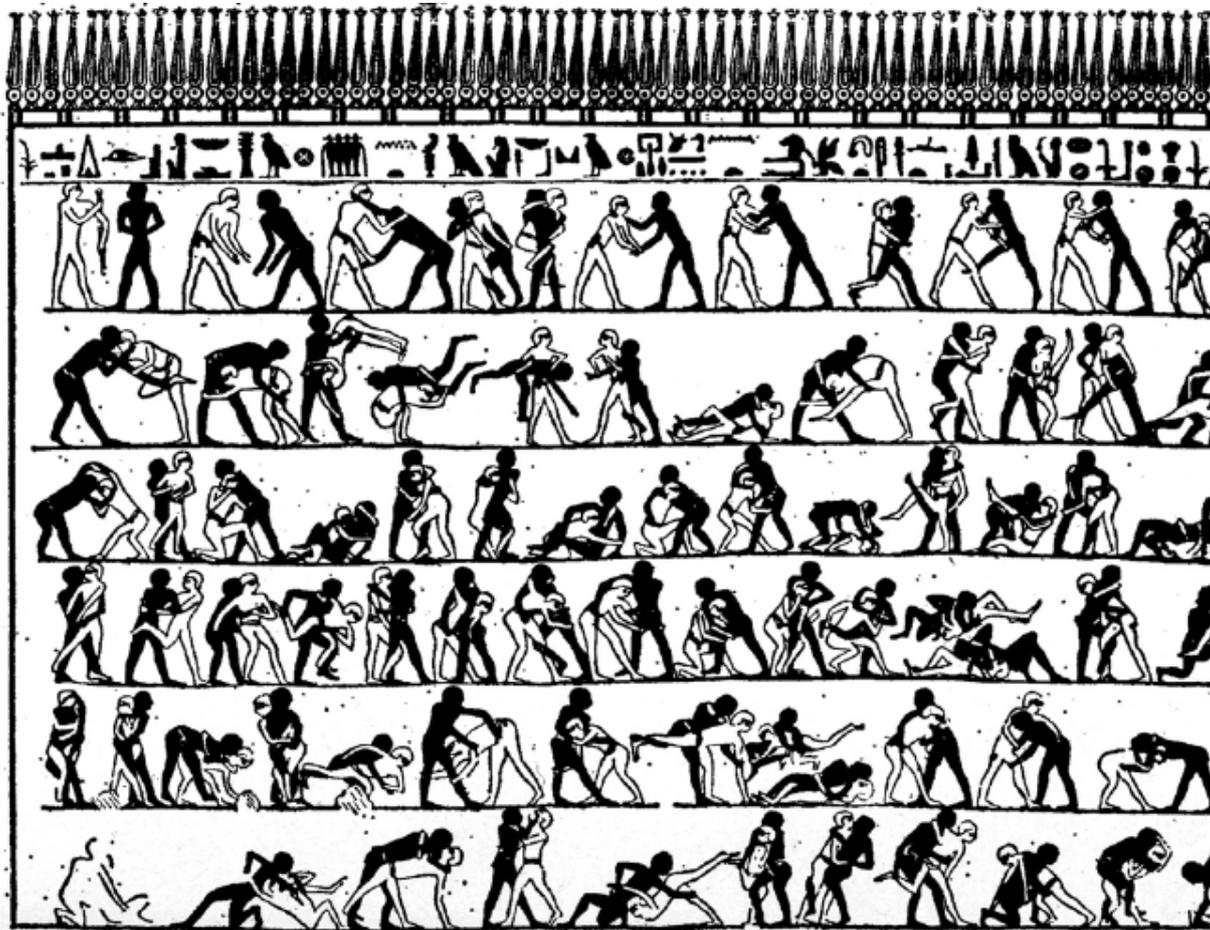
Before Animation



Shahr-e Sukhteh, Iran 3200 BCE



Before Animation



Tomb of Khnumhotep, Egypt 2400 BCE



The Phenakistoscope

- First systematic truly moving animation - the *phenakistoscope* (to be viewed in the mirror through the slit in the spinning disc)



[PHENAKISTOSCOPE - Tribute to Joseph Plateau - - YouTube](#)



First Film

- Used for research purposes in order to answer the question: *do horses lift all four limbs off the ground in gallop?*



[Sallie Gardner at a Gallop \(1878\) - YouTube](#)



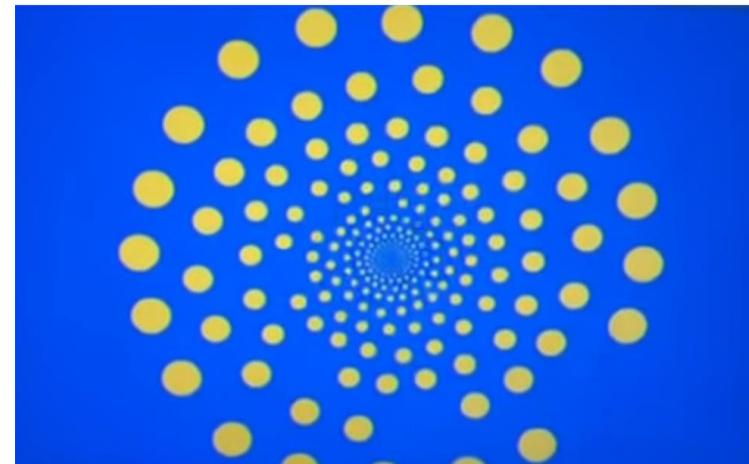
Computer animation is a sequence of still images rapidly changing at a fixed rate

The mechanism:

- Retinal persistence (our light receptors hold the perceived state over a couple of milliseconds)
scientifically disproved
- **Beta phenomenon:** visual memory in brain - not eyeball
- **Phi phenomenon:** brain anticipates, giving sense of motion (it's Gestalt psychology again!)



[Animation basics: The optical illusion of motion - TED-Ed - YouTube](#)



[Phi Phenomenon - YouTube](#)



Motion

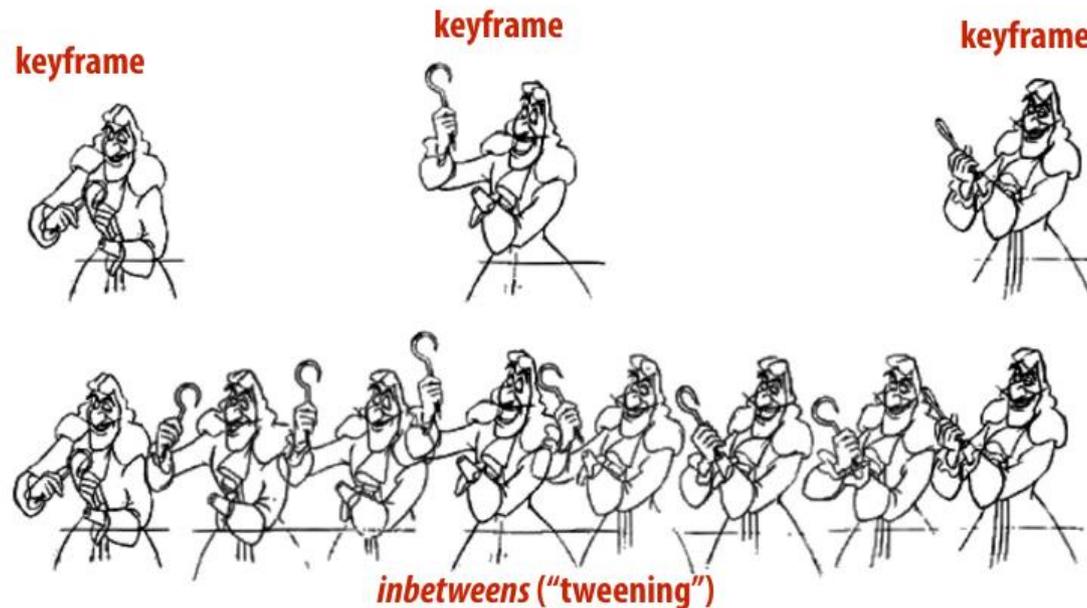
- Motion is a pre-attentive phenomenon
 - → It has a stronger power to render things distinguishable for us than color, shape, ...
- Back to Human Visual System: our eyes are more sensitive to motion at periphery
 - That's why we are prone to see “ghosts” in the corner of our visual field
- Motion triggers the orienting response / reflex (an organism's immediate response to a change in its environment, when that change is not sudden enough to elicit the startle reflex)
- Motion parallax provide 3-D cue (like stereopsis) – it means that we can understand depth in moving scenes despite not having the stereo-visual observation





“The Disney workflow”

- Senior artist draws keyframes
- Assistant draws in-betweens (tedious and labor intensive process)



In modern animation software the workflow is similar

- You, as an artist decide on the key moments of the movement, and the software interpolates the geometry in the timesteps in between

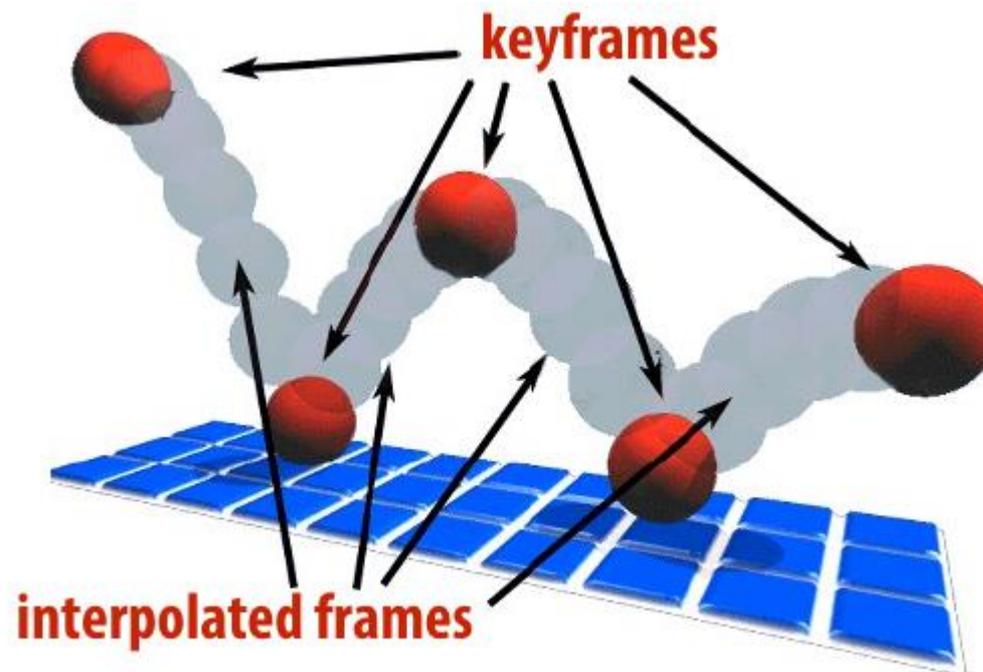


Basic idea:

- Specify important events only
- Fills in the rest via interpolation / approximation

Key frames / Events:

- Position
- Color
- Light intensity
- Camera zoom
- *etc.*





Camera

- Position
- Direction
- Focal length

Light Source

- Position
- Direction
- Radiant Power

Geometry

- Position
- Affine Transform
 - Rotation
 - Motion
 - Scaling
 - Shearing

Shading

- Transparency
- Textures
- Diffuse properties
- *etc.*

Example

- **Position** is one of the most common characteristics, which is provided via **Vec3f** values
- If the sequence contains 240 frames, for object **A** we can assign *e.g.* frames 0, 100 and 240 as *keyframes* and for object **B** - frames 10, 20 and 200
- Next we need to provide 3 positions for object **A** and 3 positions for object **B** for every keyframe, *e.g.*
 - `A.pos1 = Vec3f(7, 0, 1);`
`A.pos2 = Vec3f(10, 0, 10);`
- For the frames lying in-between 0 and 100, interpolate the position of object **A** using **A.pos1** and **A.pos2**
- By analogy proceed with object **B** and all other frames



Curve descriptions

- Explicit:

- $y = f(x)$

- $y(x) = \pm\sqrt{r^2 - x^2}$

restricted domain

- Implicit:

- $F(x, y) = 0$

- $x^2 + y^2 - r^2 = 0$

unknown solution set

- Parametric:

- $x = f_x(t), y = f_y(t)$

- $\begin{aligned} x(t) &= r \cos 2\pi t \\ y(t) &= r \sin 2\pi t \end{aligned} \quad t \in [0, 1]$

flexibility and ease of use

Polynomials

- $x(t) = a_0 + a_1t + a_2t^2 + a_3t^3 + \dots$
- Avoids complicated functions (*e.g.* `pow()`, `exp()`, `sin()`, `sqrt()`)
- Use simple polynomials of low degree



Monomial basis

- Simple basis: $1, t, t^2, \dots$ (t usually in $[0, 1]$)

Polynomial representation

$$\begin{array}{l}
 x(t) = a_0 + a_1t + a_2t^2 + a_3t^3 + \dots \\
 y(t) = b_0 + b_1t + b_2t^2 + b_3t^3 + \dots \\
 z(t) = c_0 + c_1t + c_2t^2 + c_3t^3 + \dots
 \end{array}
 \begin{array}{l}
 \text{Degree} \leftarrow \\
 \text{Coefficients } p_i \in \mathbb{R}^3 \\
 \text{Monomials} \leftarrow
 \end{array}
 \quad
 P(t) = \begin{pmatrix} x(t) \\ y(t) \\ z(t) \end{pmatrix} = \sum_{i=0}^n \begin{pmatrix} a_i \\ b_i \\ c_i \end{pmatrix} t^i$$

- Coefficients can be determined from a sufficient number of constraints (e.g. interpolation of given points)
- Given $(n + 1)$ parameter values t_i and points P_i
- Solution of a linear system in the A_i - possible, but inconvenient

Matrix representation

$$P(t)^T = (t^n \quad t^{n-1} \quad \dots \quad t \quad 1) \begin{pmatrix} a_n & b_n & c_n \\ a_{n-1} & b_{n-1} & c_{n-1} \\ \vdots & \vdots & \vdots \\ a_0 & b_0 & c_0 \end{pmatrix}$$



Derivative

- Polynomial of degree $(n - 1)$

$$\frac{dP(t)}{dt} = P'(t) = (nt^{n-1} \quad (n-1)t^{n-2} \quad \dots \quad 1 \quad 0) \begin{pmatrix} a_n & b_n & c_n \\ a_{n-1} & b_{n-1} & c_{n-1} \\ \vdots & \vdots & \vdots \\ a_0 & b_0 & c_0 \end{pmatrix}$$

- Derivative at a point is equal to the tangent vector at that point

Example

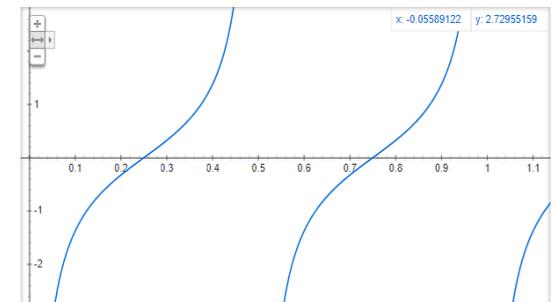
$$P(t) = (\cos 2\pi t \quad \sin 2\pi t) \begin{pmatrix} r & 0 \\ 0 & r \end{pmatrix}$$

$$P'(t) = (-2\pi \cdot \sin 2\pi t \quad 2\pi \cdot \cos 2\pi t) \begin{pmatrix} r & 0 \\ 0 & r \end{pmatrix}$$

$$x'(t) = -2\pi r \cdot \sin 2\pi t$$

$$y'(t) = 2\pi r \cdot \cos 2\pi t$$

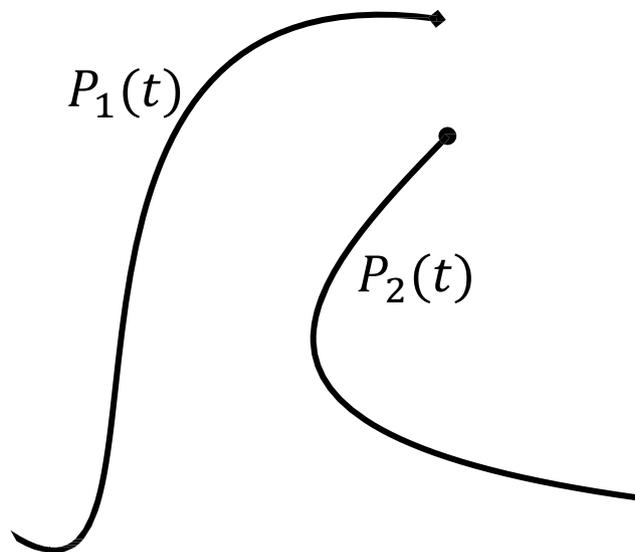
$$\frac{dy}{dx} = \frac{y'(t)}{x'(t)} = \frac{2\pi r \cdot \cos 2\pi t}{-2\pi r \cdot \sin 2\pi t} = -\operatorname{ctg} 2\pi t$$



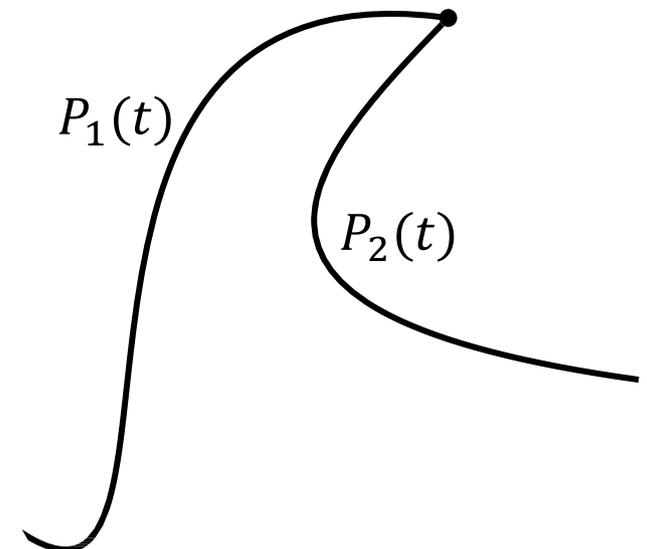


Continuity and smoothness between parametric curves

- There are two criteria for continuity:
 - Geometric continuity G^0
 - Parametric continuity C^0
- If curve P_1 ends in the same point where curve P_2 starts, it is said that we have both G^0 and C^0 continuity



Not continuous
 $P_1(1) \neq P_2(0)$

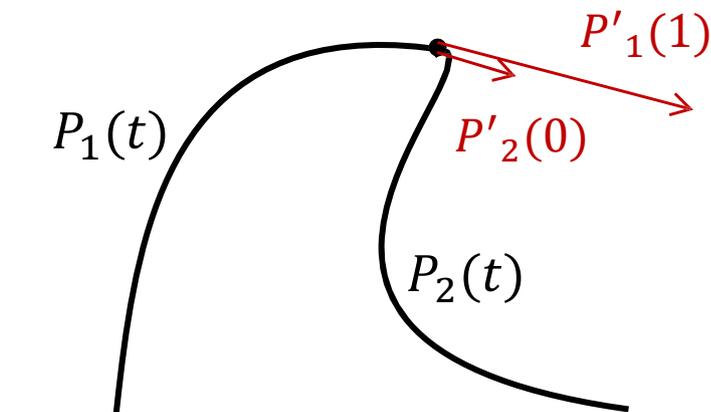


Continuous
 $P_1(1) = P_2(0)$

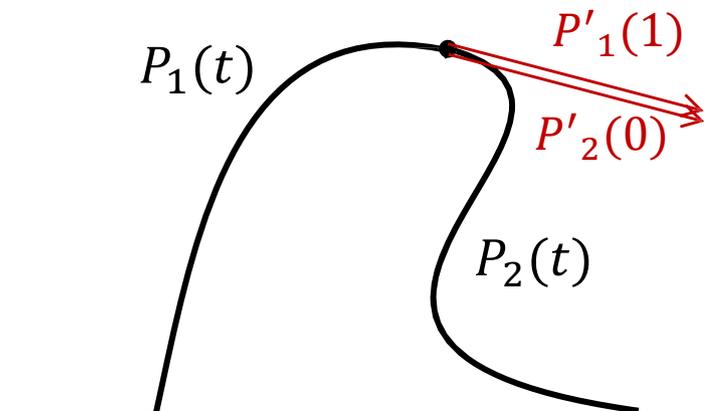


Continuity and smoothness between parametric curves

- If the tangent vectors at the joint are equally directed $P'_1(1) = kP'_2(0)$
 - It is said that we have geometric continuity G^1
- If the tangent vectors at the joint are equal $P'_1(1) = P'_2(0)$
 - It is said that we have parametric continuity C^1
- Similar for higher derivatives



G^1 -continuous
 G^0 + tangent vectors parallel
 $P'_1(1) = kP'_2(0)$



C^1 -continuous
 C^0 + tangent vectors parallel
 $P'_1(1) = P'_2(0)$

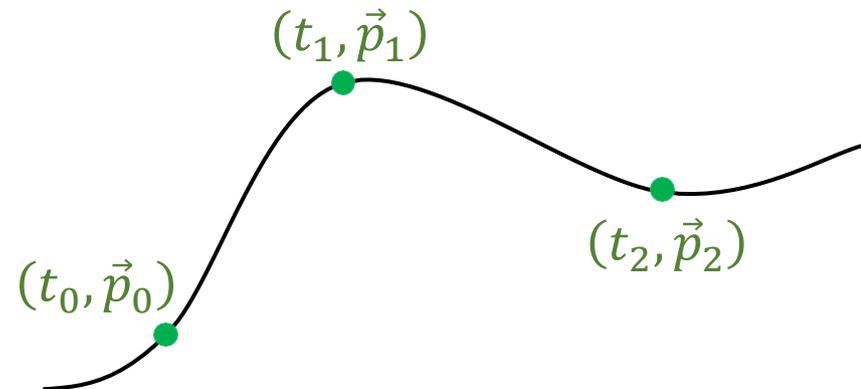


Given a set of key-points:

- (t_i, \vec{p}_i) , $t_i \in \mathbb{R}$, $\vec{p}_i \in \mathbb{R}^d$

Find a polynomial P such that:

- $\forall i P(t_i) = \vec{p}_i$





Given a set of points:

- (t_i, \vec{p}_i) , $t_i \in \mathbb{R}$, $\vec{p}_i \in \mathbb{R}^d$

Find a polynomial P such that:

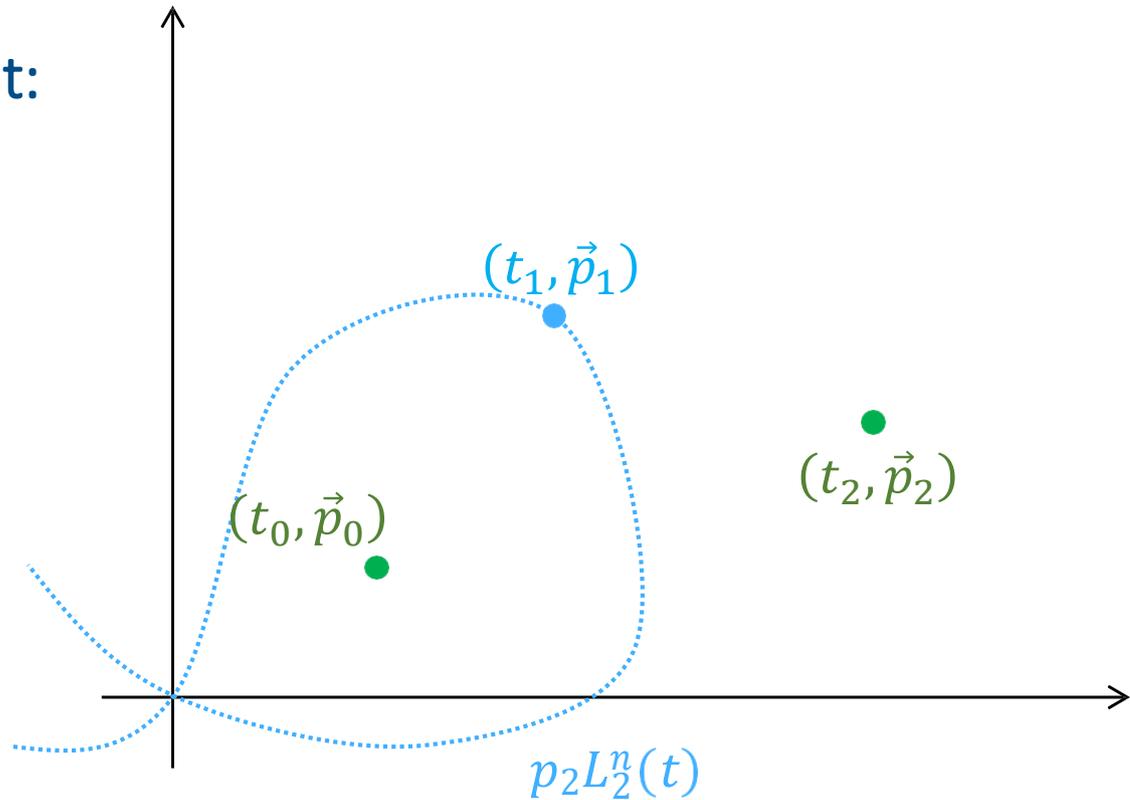
- $\forall i P(t_i) = \vec{p}_i$

For each point associate a
Lagrange basis polynomial:

$$L_i^n(t) = \prod_{\substack{j=0 \\ i \neq j}}^n \frac{t - t_j}{t_i - t_j}$$

where

$$L_i^n(t_j) = \delta_{ij} = \begin{cases} 1 & i = j \\ 0 & \text{otherwise} \end{cases}$$





Given a set of points:

- (t_i, \vec{p}_i) , $t_i \in \mathbb{R}$, $\vec{p}_i \in \mathbb{R}^d$

Find a polynomial P such that:

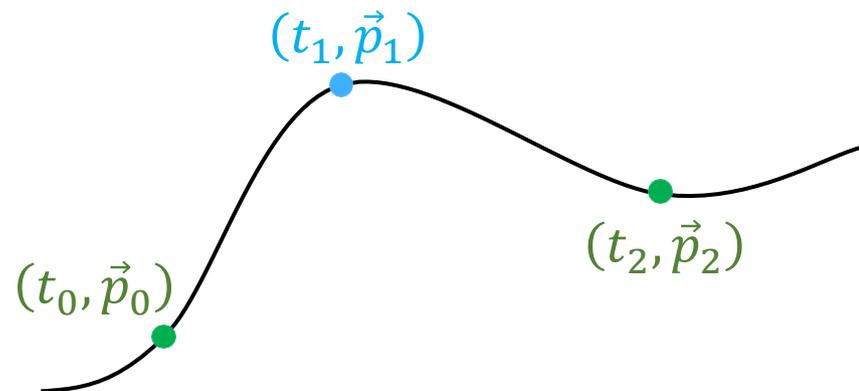
- $\forall i P(t_i) = \vec{p}_i$

For each point associate a
Lagrange basis polynomial:

$$L_i^n(t) = \prod_{\substack{j=0 \\ i \neq j}}^n \frac{t - t_j}{t_i - t_j}$$

where

$$L_i^n(t_j) = \delta_{ij} = \begin{cases} 1 & i = j \\ 0 & \text{otherwise} \end{cases}$$



Add the Lagrange basis with points as weights:

$$P(t) = \sum_{i=0}^n L_i^n(t) \vec{p}_i$$

$$P(t)^T = (L_0^n \ L_1^n \ \dots \ L_{n-1}^n) \begin{pmatrix} p_{0,x} & p_{0,y} & p_{0,z} \\ p_{1,x} & p_{1,y} & p_{1,z} \\ \vdots & \vdots & \vdots \\ p_{n-1,x} & p_{n-1,y} & p_{n-1,z} \end{pmatrix}$$



For each point associate a Lagrange basis polynomial:

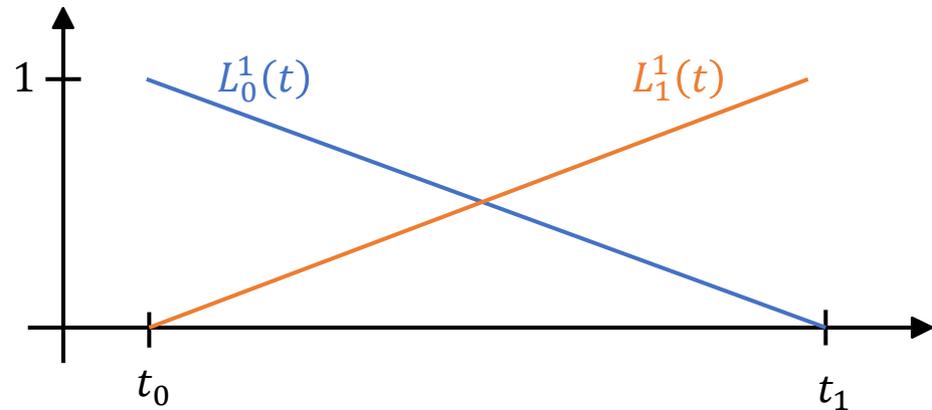
$$L_i^n(t) = \prod_{\substack{j=0 \\ i \neq j}}^n \frac{t - t_j}{t_i - t_j}$$

Simple Linear Interpolation

- $T = \{t_0, t_1\}$

$$L_0^1(t) = \frac{t - t_1}{t_0 - t_1}$$

$$L_1^1(t) = \frac{t - t_0}{t_1 - t_0}$$



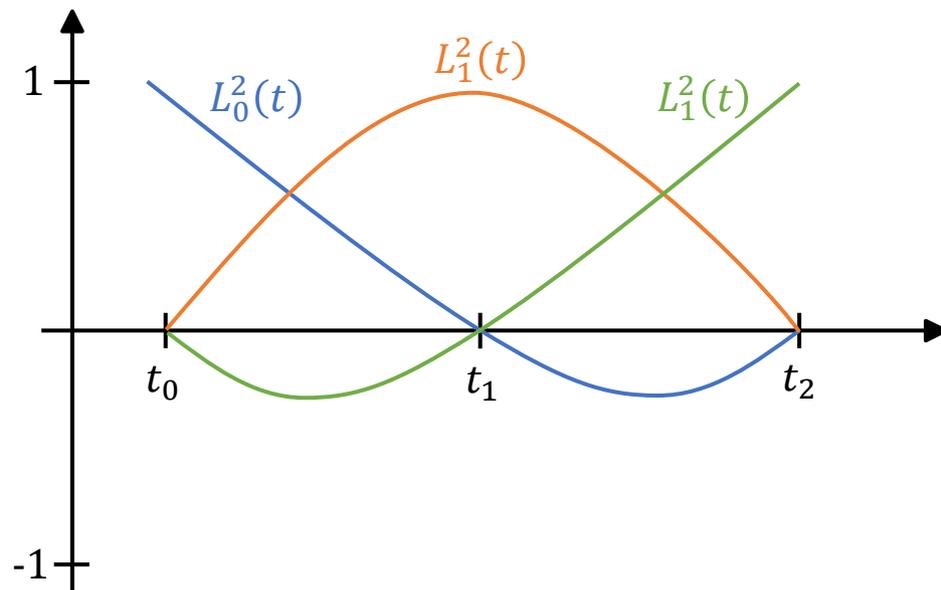
Simple Quadratic Interpolation

- $T = \{t_0, t_1, t_2\}$

$$L_0^2(t) = \frac{t - t_1}{t_0 - t_1} \frac{t - t_2}{t_0 - t_2}$$

$$L_1^2(t) = \frac{t - t_0}{t_1 - t_0} \frac{t - t_2}{t_1 - t_2}$$

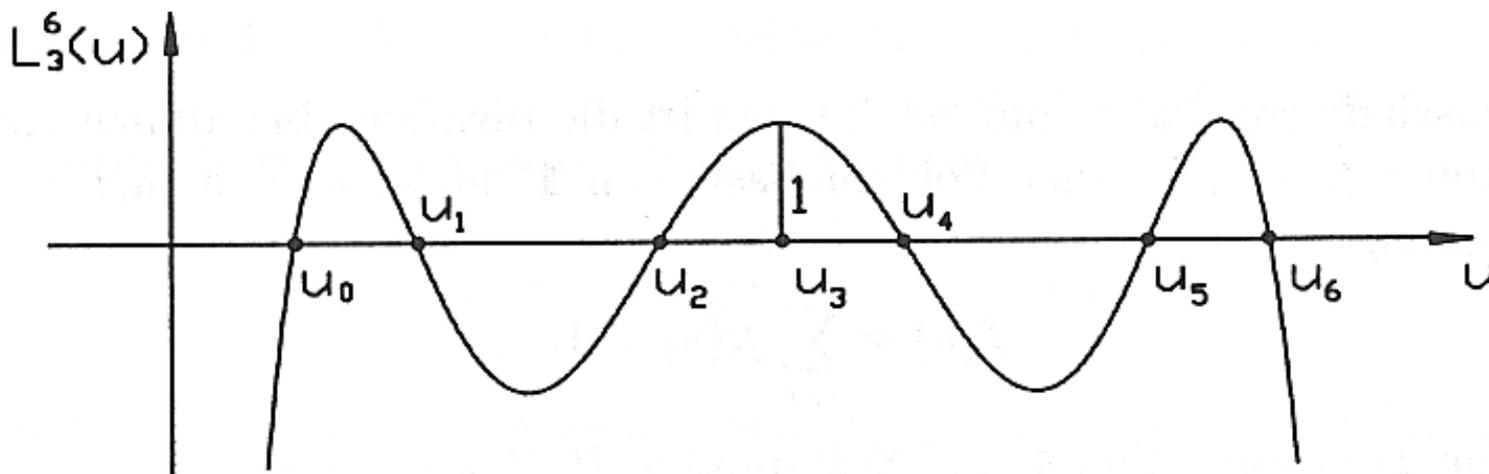
$$L_2^2(t) = \frac{t - t_0}{t_2 - t_0} \frac{t - t_1}{t_2 - t_1}$$





Problems with a single polynomial

- Degree depends on the number of interpolation constraints
- Strong overshooting for high degree ($n > 7$)
- Problems with smooth joints
- Numerically unstable
- No local changes



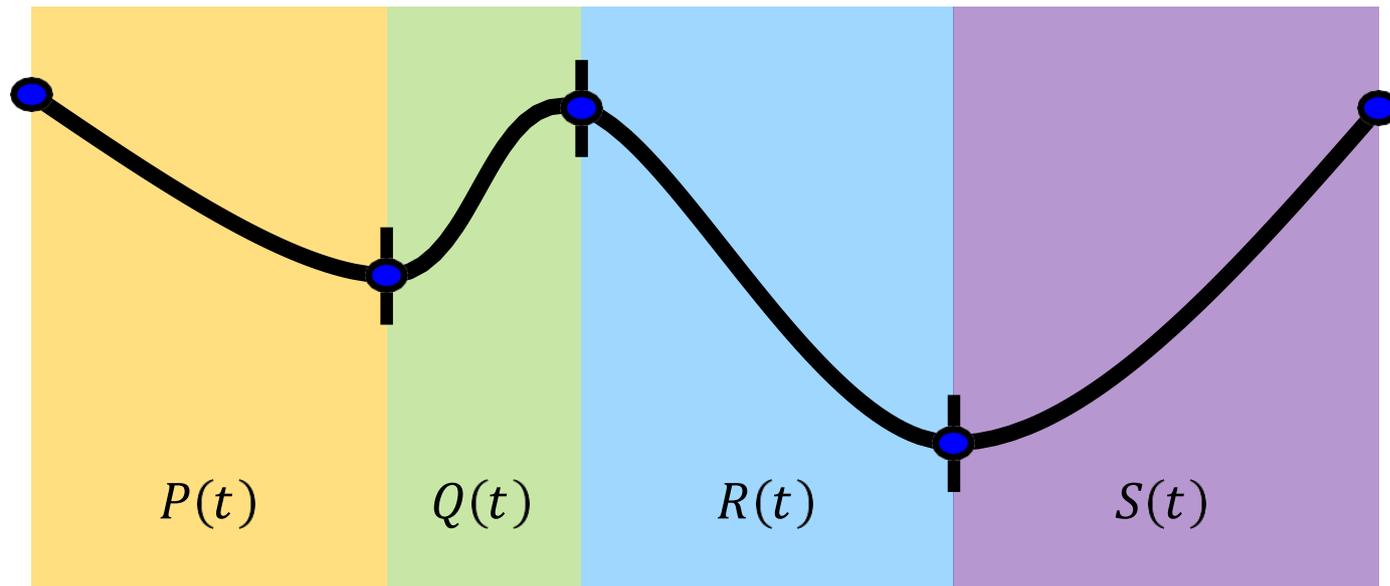


Functions for interpolation & approximation

- Standard curve and surface primitives in geometric modeling
- Key frame and in-betweens in animations
- Filtering and reconstruction of images

Historically

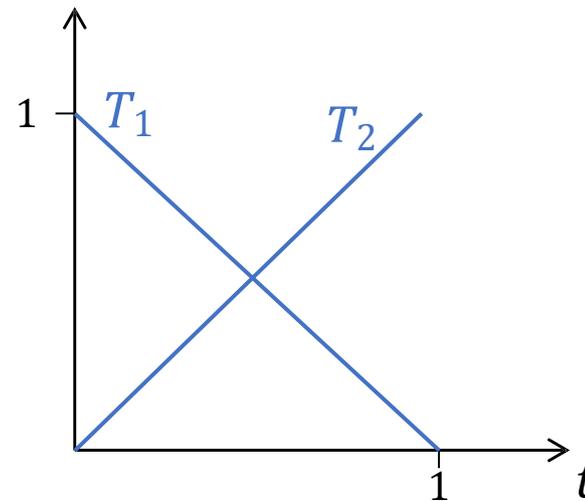
- Name for a tool in ship building
 - Flexible metal strip that tries to stay straight
- Within computer graphics:
 - Piecewise polynomial function





Linear splines

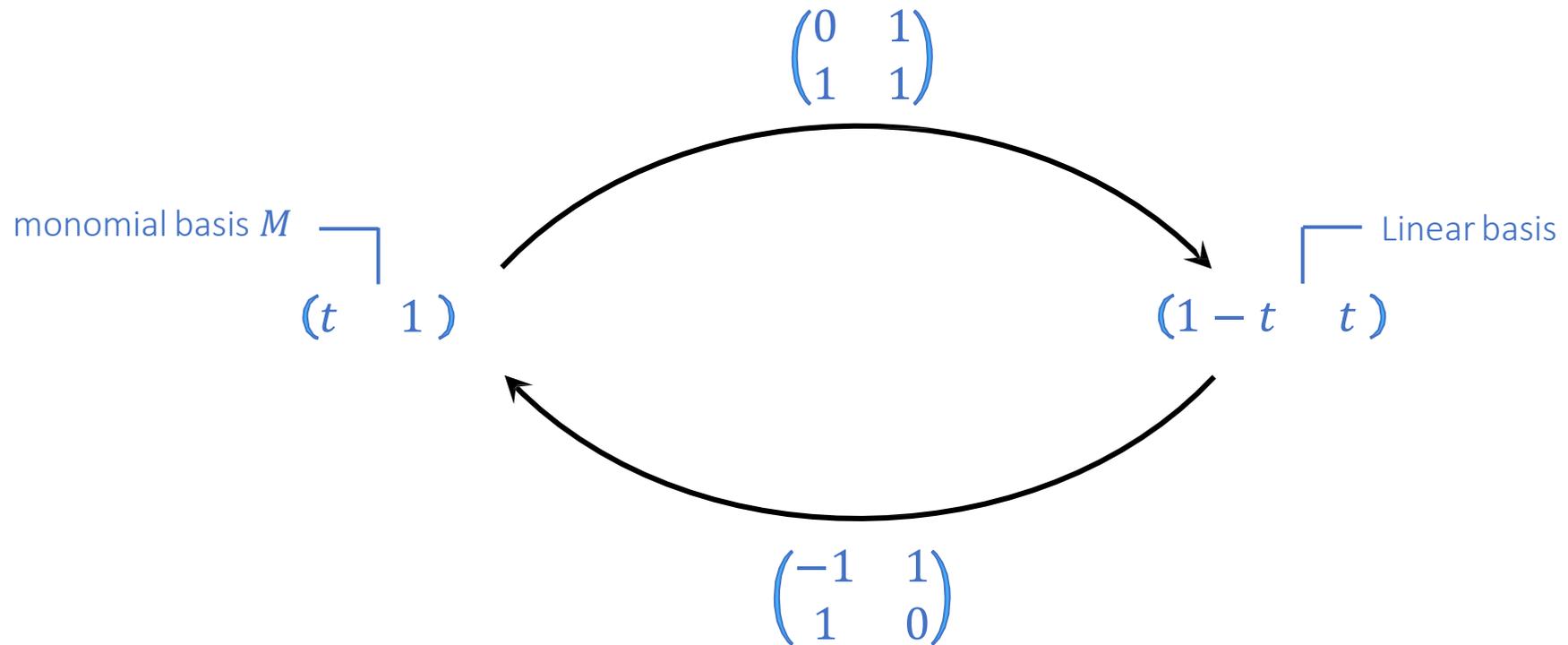
- Defined by two points: \vec{p}_1, \vec{p}_2
- Searching for $P(t)$ such that:
 - $P(0) = \vec{p}_1$
 - $P(1) = \vec{p}_2$
 - Degree of P is 1
- Basis:
 - $T_1(t) = 1 - t$
 - $T_2(t) = t$



$$P(t) = \vec{p}_1 T_1(t) + \vec{p}_2 T_2(t)$$

Linear basis

$$P(t)^T = \begin{pmatrix} 1 - t & t \end{pmatrix} \begin{pmatrix} \vec{p}_1^T \\ \vec{p}_2^T \end{pmatrix}$$

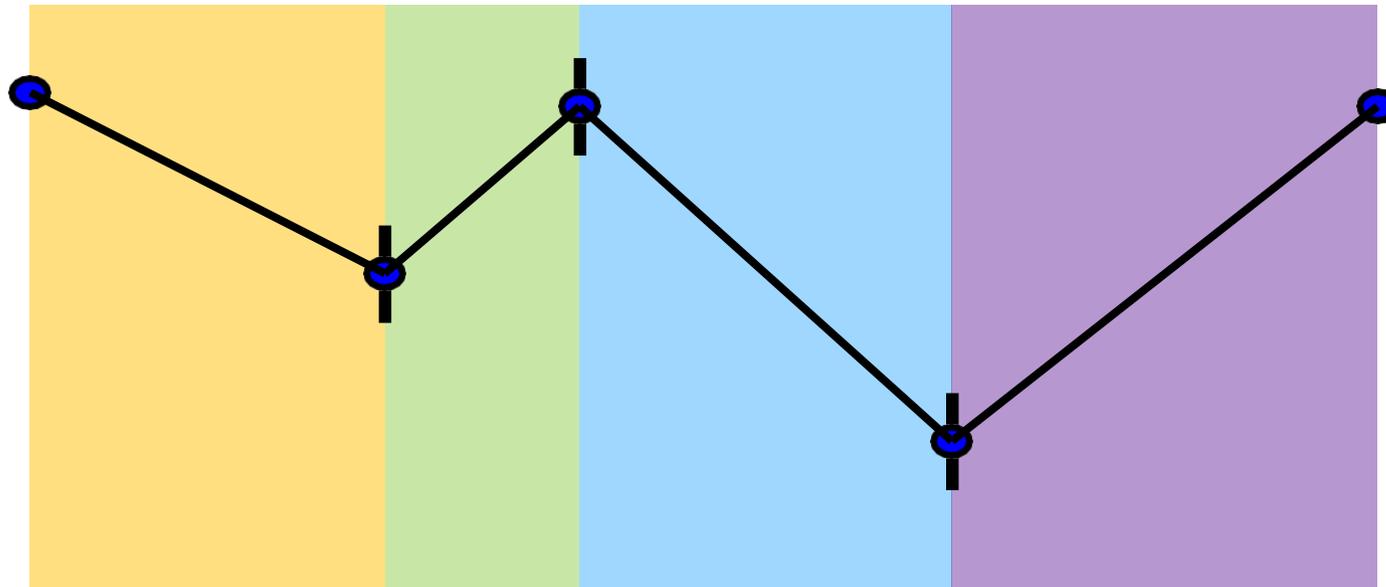


$$P(t)^T = M \cdot \begin{pmatrix} -1 & 1 \\ 1 & 0 \end{pmatrix} \cdot \begin{pmatrix} \vec{p}_1^T \\ \vec{p}_2^T \end{pmatrix}$$



$$P(t)^T = M \cdot \begin{pmatrix} -1 & 1 \\ 1 & 0 \end{pmatrix} \cdot \begin{pmatrix} \vec{p}_1^T \\ \vec{p}_2^T \end{pmatrix}$$

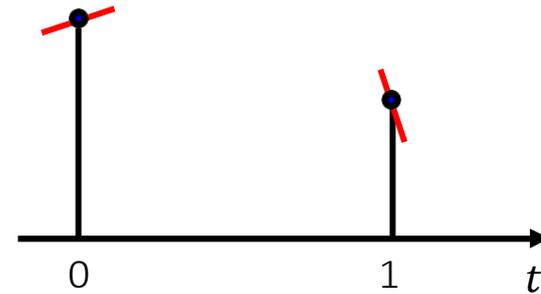
C^0 -continuous





Cubic splines

- Defined by two points: \vec{p}_1, \vec{p}_2 and two tangents: $\vec{\tau}_1, \vec{\tau}_2$
- Searching for $P(t)$ such that:
 - $P(0) = \vec{p}_1$
 - $P'(0) = \vec{\tau}_1$
 - $P'(1) = \vec{\tau}_2$
 - $P(1) = \vec{p}_2$
 - Degree of P is 3
- Basis:
 - $H_0^3(t) = ?$
 - $H_1^3(t) = ?$
 - $H_2^3(t) = ?$
 - $H_3^3(t) = ?$



$$P(t) = P(0)H_0^3(t) + P'(0)H_1^3(t) + P'(1)H_2^3(t) + P(1)H_3^3(t)$$



Cubic splines

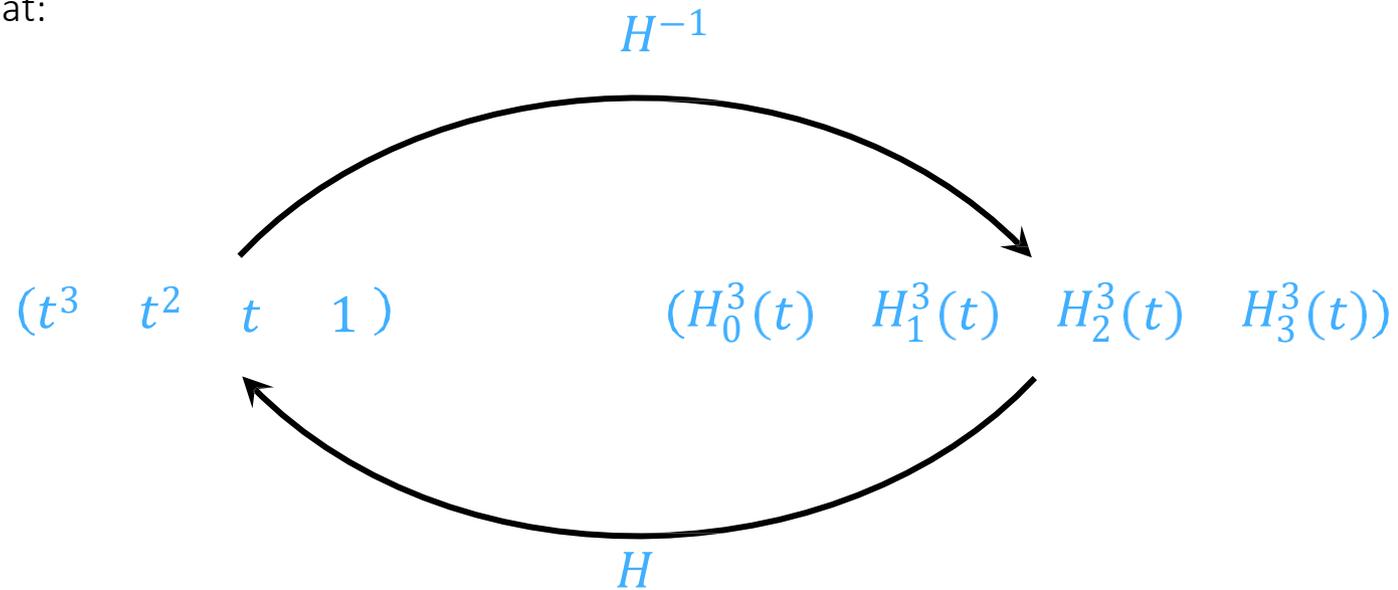
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- Searching for $P(t)$ such that:

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- $P'(0) = \vec{\tau}_1$
- $P'(1) = \vec{\tau}_2$
- $P(1) = \vec{p}_2$
- Degree of P is 3

- Basis:

- $H_0^3(t) = ?$
- $H_1^3(t) = ?$
- $H_2^3(t) = ?$
- $H_3^3(t) = ?$



$$P(t)^\top = M \cdot H \cdot \begin{pmatrix} \vec{p}_1^\top \\ \vec{\tau}_1^\top \\ \vec{\tau}_2^\top \\ \vec{p}_2^\top \end{pmatrix} = M \cdot H \cdot G$$



Cubic splines

- Defined by two points: \vec{p}_1, \vec{p}_2 and two tangents: $\vec{\tau}_1, \vec{\tau}_2$

- Searching for $P(t)$ such that:

- $P(0) = \vec{p}_1$
- $P'(0) = \vec{\tau}_1$
- $P'(1) = \vec{\tau}_2$
- $P(1) = \vec{p}_2$
- Degree of P is 3

- Basis:

- $H_0^3(t) = ?$
- $H_1^3(t) = ?$
- $H_2^3(t) = ?$
- $H_3^3(t) = ?$

- $P(t)^\top = (t^3 \ t^2 \ t \ 1) \cdot H \cdot G$
- $P'(t)^\top = (3t^2 \ 2t \ 1 \ 0) \cdot H \cdot G$

- $\vec{p}_1^\top = P(0)^\top = (0 \ 0 \ 0 \ 1) \cdot H \cdot G$
- $\vec{\tau}_1^\top = P'(0)^\top = (0 \ 0 \ 1 \ 0) \cdot H \cdot G$
- $\vec{\tau}_2^\top = P'(1)^\top = (3 \ 2 \ 1 \ 0) \cdot H \cdot G$
- $\vec{p}_2^\top = P(1)^\top = (1 \ 1 \ 1 \ 1) \cdot H \cdot G$

$$\begin{pmatrix} \vec{p}_1^\top \\ \vec{\tau}_1^\top \\ \vec{\tau}_2^\top \\ \vec{p}_2^\top \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 3 & 2 & 1 & 0 \\ 1 & 1 & 1 & 1 \end{pmatrix} \cdot H \cdot \begin{pmatrix} \vec{p}_1^\top \\ \vec{\tau}_1^\top \\ \vec{\tau}_2^\top \\ \vec{p}_2^\top \end{pmatrix}$$



Cubic splines

- Defined by two points: \vec{p}_1, \vec{p}_2 and two tangents: $\vec{\tau}_1, \vec{\tau}_2$
- Searching for $P(t)$ such that:
 - $P(0) = \vec{p}_1$
 - $P'(0) = \vec{\tau}_1$
 - $P'(1) = \vec{\tau}_2$
 - $P(1) = \vec{p}_2$
 - Degree of P is 3
- Basis:
 - $H_0^3(t) = ?$
 - $H_1^3(t) = ?$
 - $H_2^3(t) = ?$
 - $H_3^3(t) = ?$

$$H = \begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 3 & 2 & 1 & 0 \\ 1 & 1 & 1 & 1 \end{pmatrix}^{-1} = \begin{pmatrix} 2 & 1 & 1 & -2 \\ -3 & -2 & -1 & 3 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix}$$

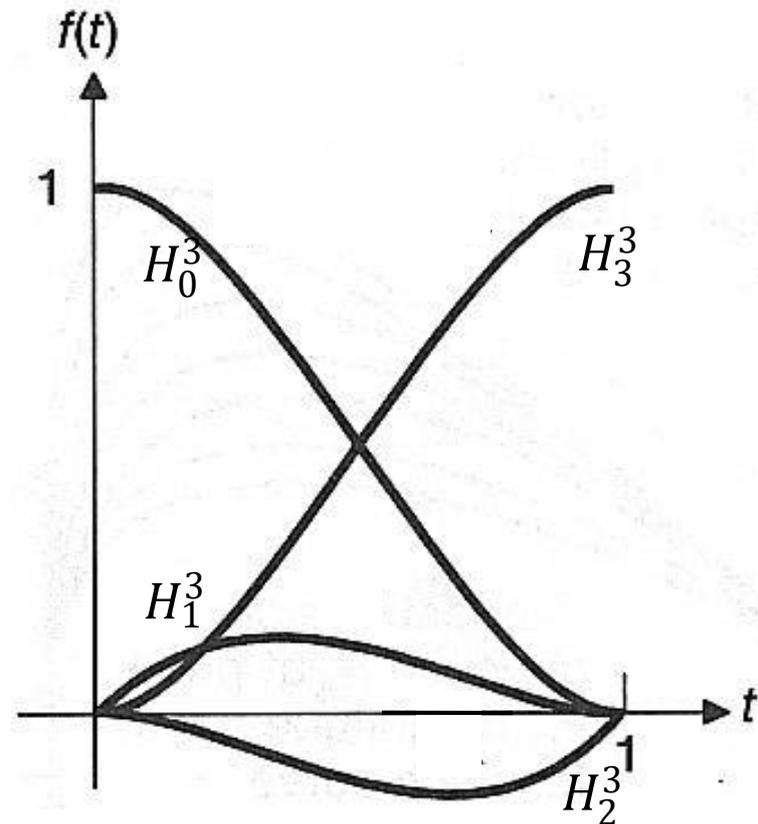


Cubic splines

- Defined by two points: \vec{p}_1, \vec{p}_2 and two tangents: $\vec{\tau}_1, \vec{\tau}_2$
- Searching for $P(t)$ such that:
 - $P(0) = \vec{p}_1$
 - $P'(0) = \vec{\tau}_1$
 - $P'(1) = \vec{\tau}_2$
 - $P(1) = \vec{p}_2$
 - Degree of P is 3
- Basis:
 - $H_0^3(t) = (1-t)^2(1+2t)$
 - $H_1^3(t) = t(1-t)^2$
 - $H_2^3(t) = t^2(t-1)$
 - $H_3^3(t) = (3-2t)t^2$

$$H = \begin{pmatrix} 2 & 1 & 1 & -2 \\ -3 & -2 & -1 & 3 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix}$$

$(H_0^3(t) \quad H_1^3(t) \quad H_2^3(t) \quad H_3^3(t))$





Cubic splines

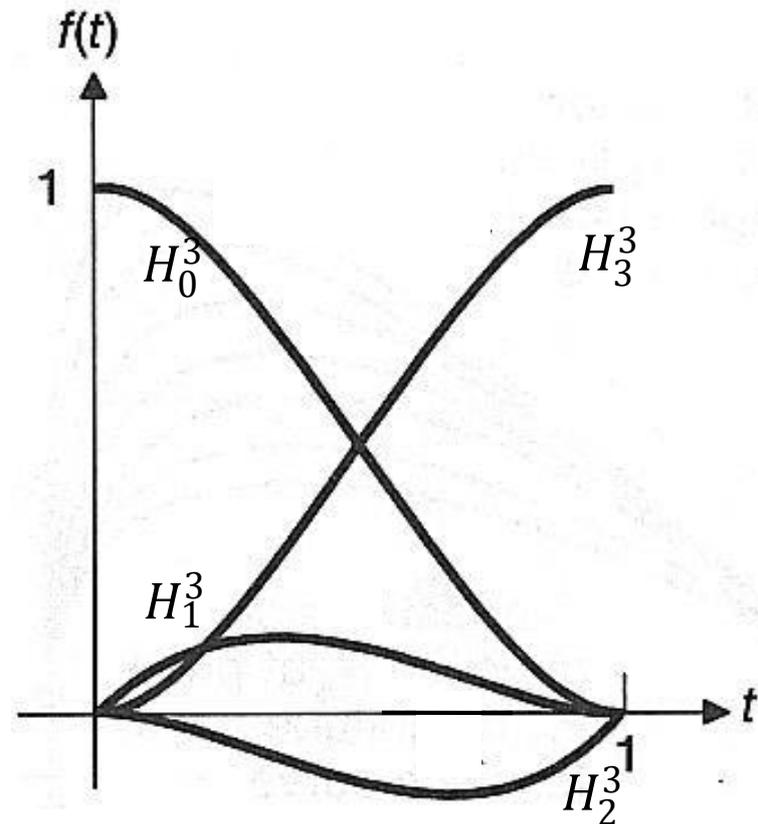
- Basis:
 - $H_0^3(t) = (1 - t)^2(1 + 2t)$
 - $H_1^3(t) = t(1 - t)^2$
 - $H_2^3(t) = t^2(t - 1)$
 - $H_3^3(t) = (3 - 2t)t^2$

$$H = \begin{pmatrix} 2 & 1 & 1 & -2 \\ -3 & -2 & -1 & 3 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix}$$

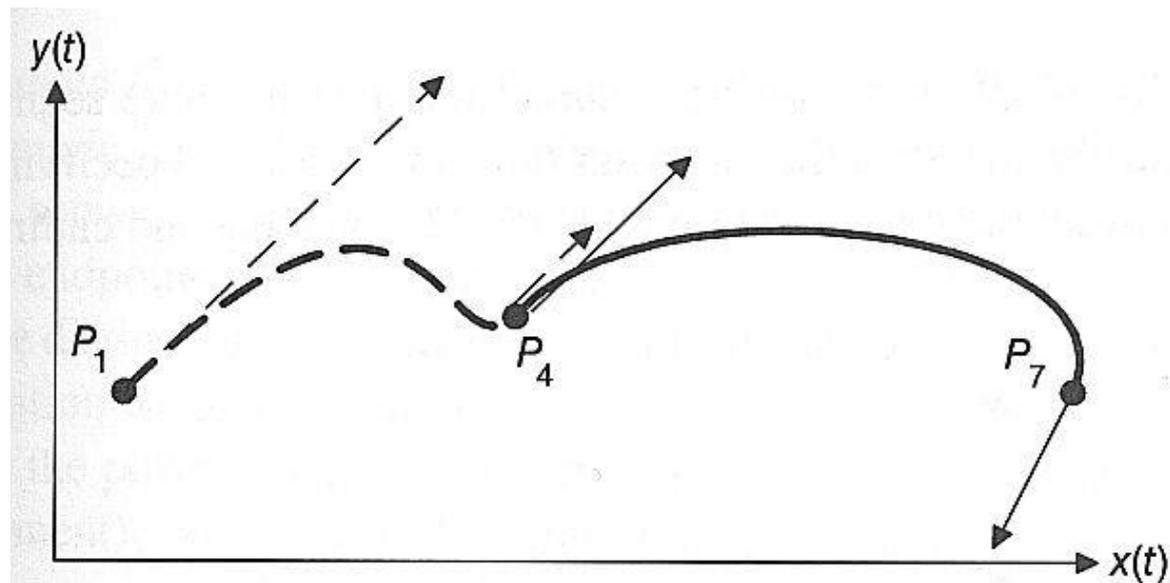
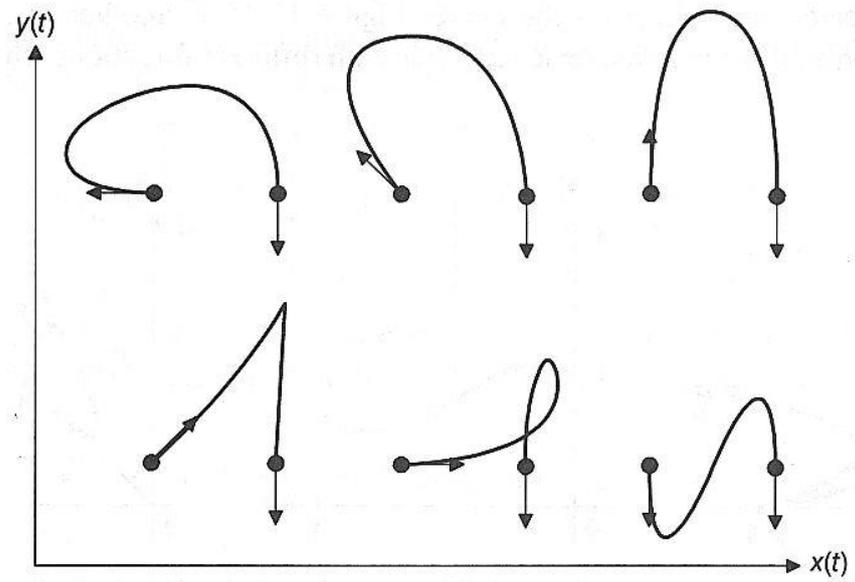
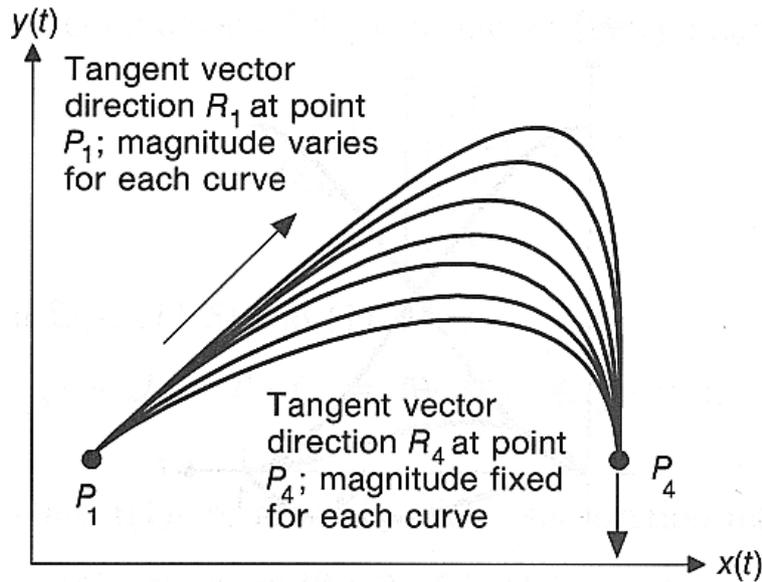
$(H_0^3(t) \quad H_1^3(t) \quad H_2^3(t) \quad H_3^3(t))$

Properties of Hermite Basis Functions

- H_0^3 (H_3^3) interpolates smoothly from 1 to 0
- H_0^3 and H_3^3 have zero derivative at $t = 0$ and $t = 1$
 - No contribution to derivative (H_1^3, H_2^3)
- H_1^3 and H_2^3 are zero at $t = 0$ and $t = 1$
 - No contribution to position (H_0^3, H_3^3)
- H_1^3 (H_2^3) has slope 1 at $t = 0$ ($t = 1$)
 - Unit factor for specified derivative vector



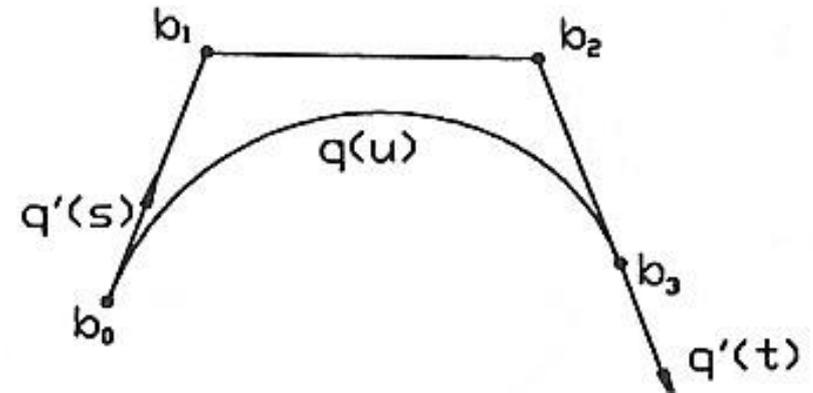
Examples: Hermite Interpolation





Bézier splines

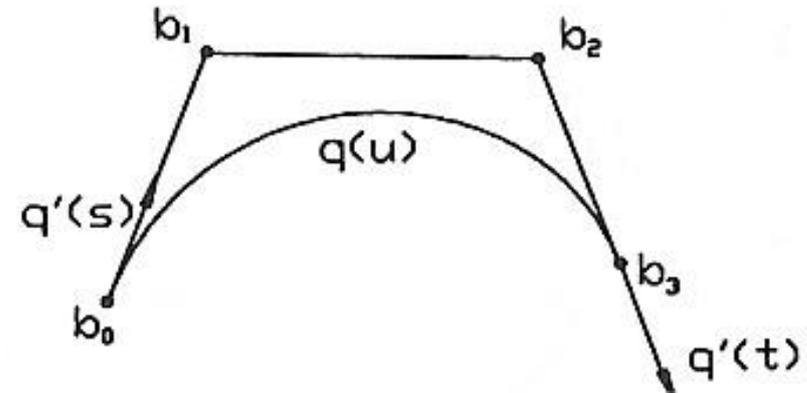
- Defined by 4 points:
 - b_0, b_3 : start and end points
 - b_1, b_2 : control points that are approximated
- Searching for $P(t)$ such that:
 - $P(0) = b_0$
 - $P'(0) = 3(b_1 - b_0)$
 - $P'(1) = 3(b_3 - b_2)$
 - $P(1) = b_3$
 - Degree of P is 3





Bézier splines

- Defined by 4 points:
 - b_0, b_3 : start and end points
 - b_1, b_2 : control points that are approximated
- Searching for $P(t)$ such that:
 - $P(0) = b_0$
 - $P'(0) = 3(b_1 - b_0)$
 - $P'(1) = 3(b_3 - b_2)$
 - $P(1) = b_3$
 - Degree of P is 3



$$\begin{pmatrix} p_1^\top \\ t_1^\top \\ t_2^\top \\ p_2^\top \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ -3 & 3 & 0 & 0 \\ 0 & 0 & -3 & 3 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} b_0^\top \\ b_1^\top \\ b_2^\top \\ b_3^\top \end{pmatrix}$$

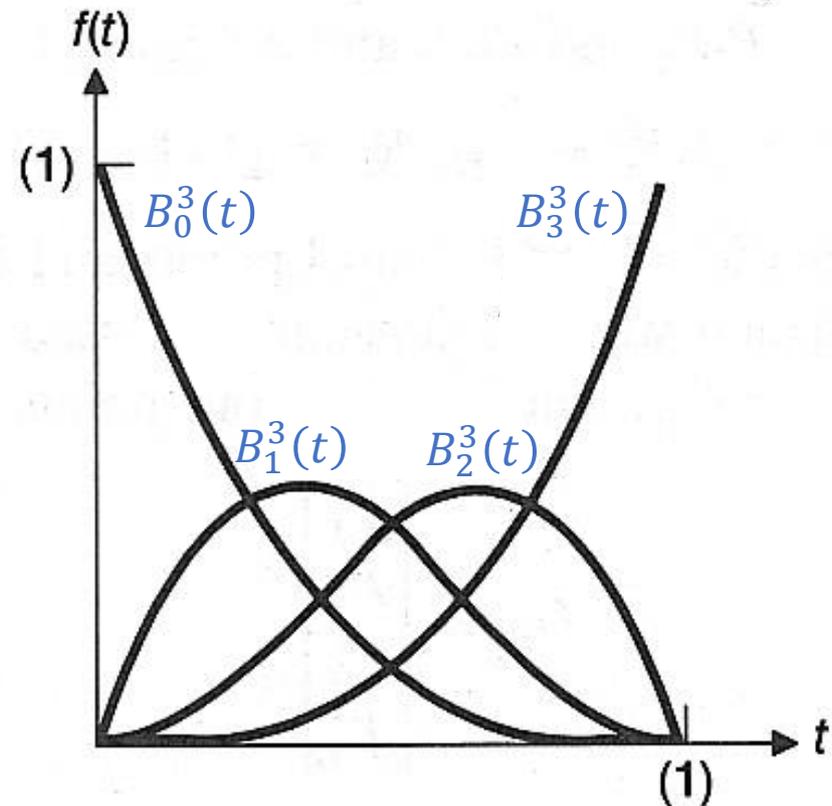
$$P(t)^\top = M \cdot H \cdot T_{BH} \cdot G$$



Bézier splines

- Defined by 4 points:
 - b_0, b_3 : start and end points
 - b_1, b_2 : control points that are approximated
- Searching for $P(t)$ such that:
 - $P(0) = b_0$
 - $P'(0) = 3(b_1 - b_0)$
 - $P'(1) = 3(b_3 - b_2)$
 - $P(1) = b_3$
 - Degree of P is 3
- Basis:
 - $B_0^3(t) = (1 - t)^3$
 - $B_1^3(t) = 3t(1 - t)^2$
 - $B_2^3(t) = 3t^2(1 - t)$
 - $B_3^3(t) = t^3$
- Bernstein polynomial:
 - $B_i^n(t) = \binom{n}{i} t^i (1 - t)^{n-i}$

$$B = H \cdot T_{BH} = \begin{pmatrix} -1 & 3 & -3 & 1 \\ 3 & -6 & 3 & 0 \\ -3 & 3 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix}$$



$$P(t) = b_0 B_0^3(t) + b_1 B_1^3(t) + b_2 B_2^3(t) + b_3 B_3^3(t)$$

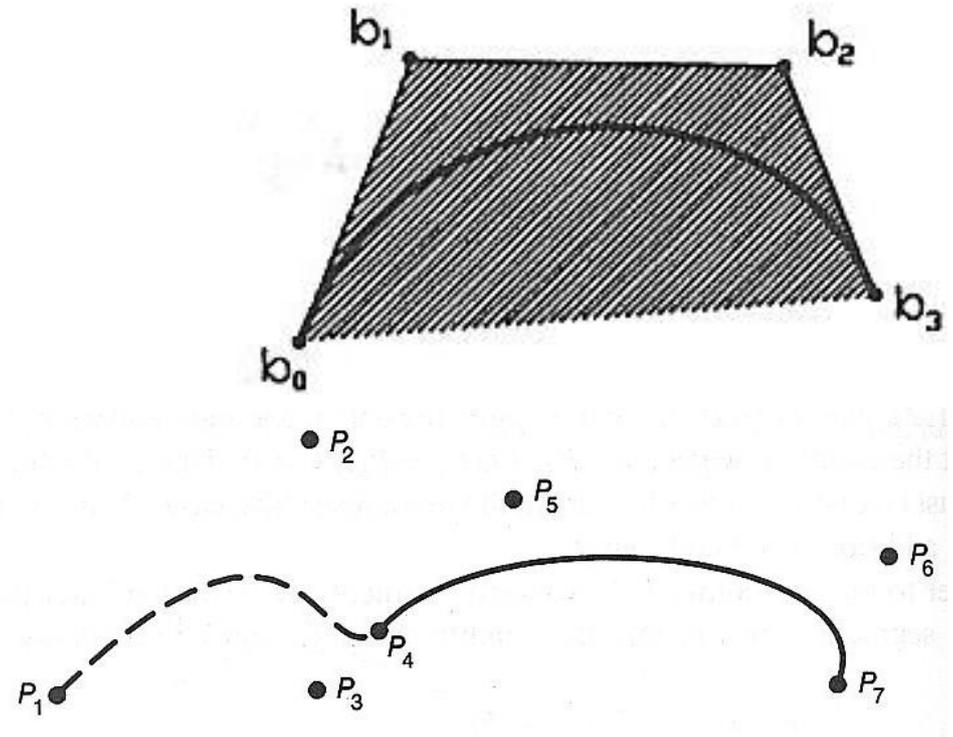


Advantages:

- End point interpolation
- Tangents explicitly specified
- Smooth joints are simple
 - P_3, P_4, P_5 collinear $\rightarrow G^1$ continuous
 - $P_5 - P_4 = P_4 - P_3 \rightarrow C^1$ continuous
- Geometric meaning of control points
- Affine invariance
- Convex hull property
 - For $0 < t < 1$: $B_i(t) \geq 0$
- Symmetry: $B_i(t) = B_{n-i}(1-t)$

Disadvantages

- Smooth joints need to be maintained explicitly
 - Automatic in B-Splines (and NURBS)





Direct evaluation of the basis functions $P(t) = \sum_i b_i B_i^n(t)$

- Simple but expensive

Use recursion

- Recursive definition of the basis functions

$$B_i^n(t) = \binom{n}{i} t^i (1-t)^{n-i} = t B_{i-1}^{n-1}(t) + (1-t) B_i^{n-1}(t)$$

- Inserting this once yields:

$$P(t) = \sum_{i=0}^n b_i^0 B_i^n(t) = \sum_{i=0}^{n-1} b_i^1 B_i^{n-1}(t)$$

- With the new Bézier points given by the recursion

$$b_i^0(t) = b_i$$

$$b_i^k(t) = t b_{i+1}^{k-1}(t) + (1-t) b_i^{k-1}(t)$$



DeCasteljau Algorithm:

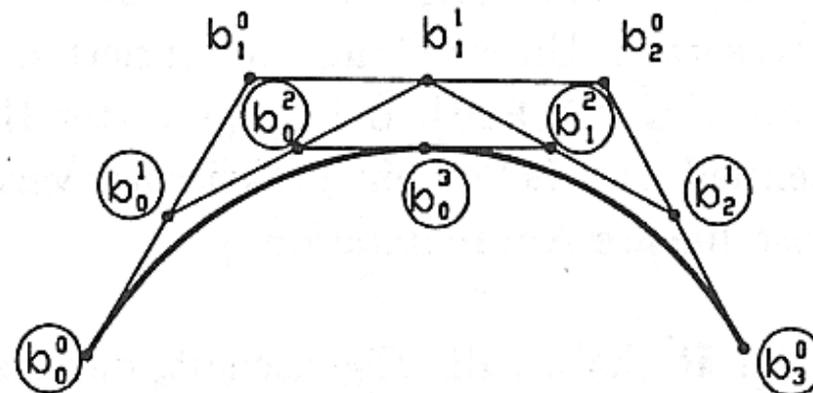
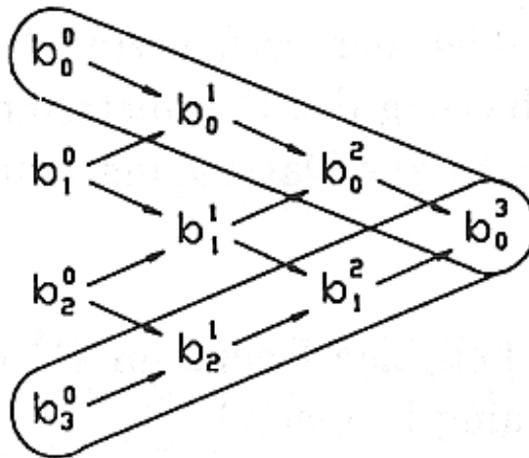
- Recursive degree reduction of the Bezier curve by using the recursion formula for the Bernstein polynomials

$$P(t) = \sum_{i=0}^n b_i^0 B_i^n(t) = \sum_{i=0}^{n-1} b_i^1 B_i^{n-1}(t) = \dots = b_i^n(t) \cdot 1$$

$$b_i^k(t) = t b_{i+1}^{k-1}(t) + (1-t) b_i^{k-1}(t)$$

Example:

- $t = 0.5$



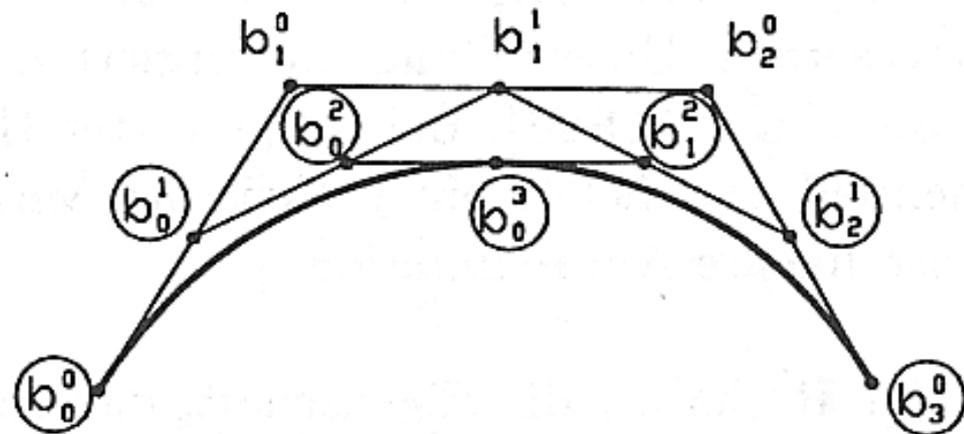
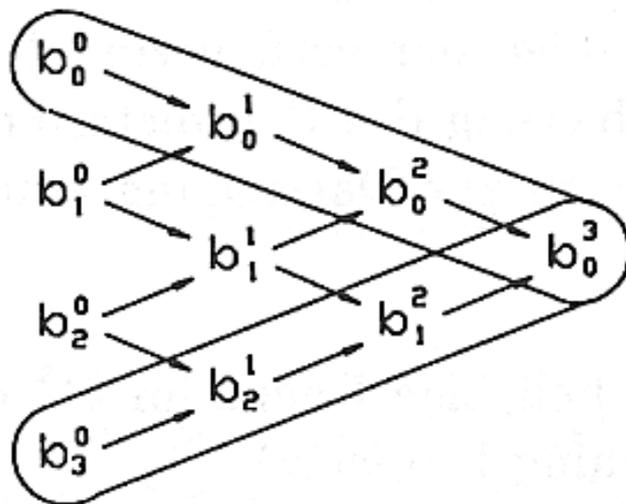


Subdivision using the deCasteljau Algorithm

- Take boundaries of the deCasteljau triangle as new control points for left / right portion of the curve

Extrapolation

- Backwards subdivision
 - Reconstruct triangle from one side



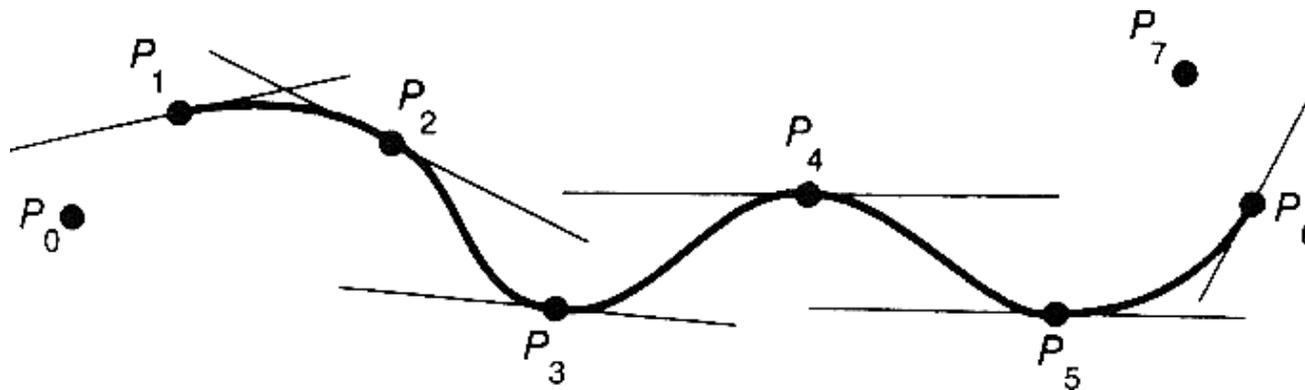


Goal

- Smooth (C^1)-joints between (cubic) spline segments

Algorithm

- Tangents given by neighboring points P_{i-1} P_{i+1}
- Construct (cubic) Hermite segments



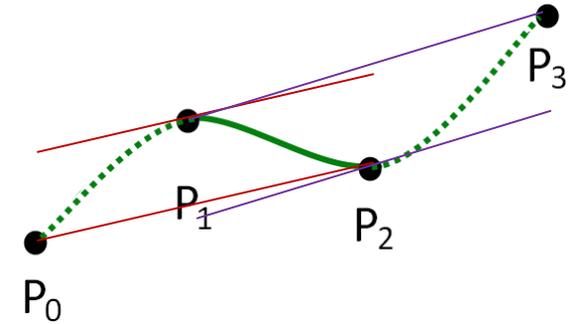
Advantage

- Arbitrary number of control points
- Interpolation without overshooting
- Local control



Catmull-Rom splines

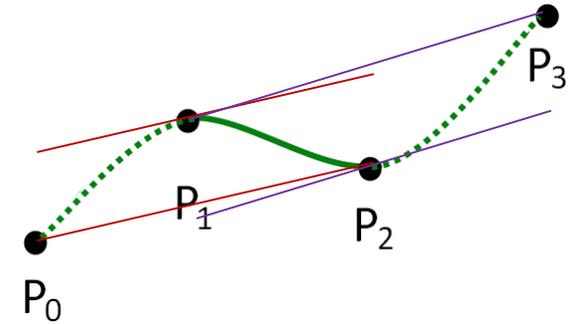
- Defined by 4 points:
 - c_1, c_2 : start and end points
 - c_0, c_3 : neighbor segment points
- Searching for $P(t)$ such that:
 - $P(0) = c_1$
 - $P'(0) = \frac{1}{2}(c_2 - c_0)$
 - $P'(1) = \frac{1}{2}(c_3 - c_1)$
 - $P(1) = c_2$
 - Degree of P is 3





Catmull-Rom splines

- Defined by 4 points:
 - c_1, c_2 : start and end points
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- Searching for $P(t)$ such that:
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 - $P(1) = c_2$
 - Degree of P is 3



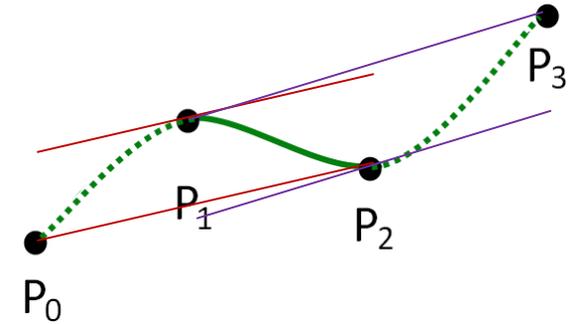
$$\begin{pmatrix} p_1^\top \\ t_1^\top \\ t_2^\top \\ p_2^\top \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ -0.5 & 0 & 0.5 & 0 \\ 0 & -0.5 & 0 & 0.5 \\ 0 & 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} c_0^\top \\ c_1^\top \\ c_2^\top \\ c_3^\top \end{pmatrix}$$

$$P(t)^\top = M \cdot H \cdot T_{CH} \cdot G$$



Catmull-Rom splines

- Defined by 4 points:
 - c_1, c_2 : start and end points
 - c_0, c_3 : neighbor segment points
- Searching for $P(t)$ such that:
 - $P(0) = c_1$
 - $P'(0) = \frac{1}{2}(c_2 - c_0)$
 - $P'(1) = \frac{1}{2}(c_3 - c_1)$
 - $P(1) = c_2$
 - Degree of P is 3
- Basis:
 - $C_0^3(t) = \frac{1}{2}t(1-t)^2$
 - $C_1^3(t) = \frac{1}{2}(t-1)(3t^2 - 2t - 2)$
 - $C_2^3(t) = -\frac{1}{2}t(3t^2 - 4t - 1)$
 - $C_3^3(t) = \frac{1}{2}t^2(t-1)$

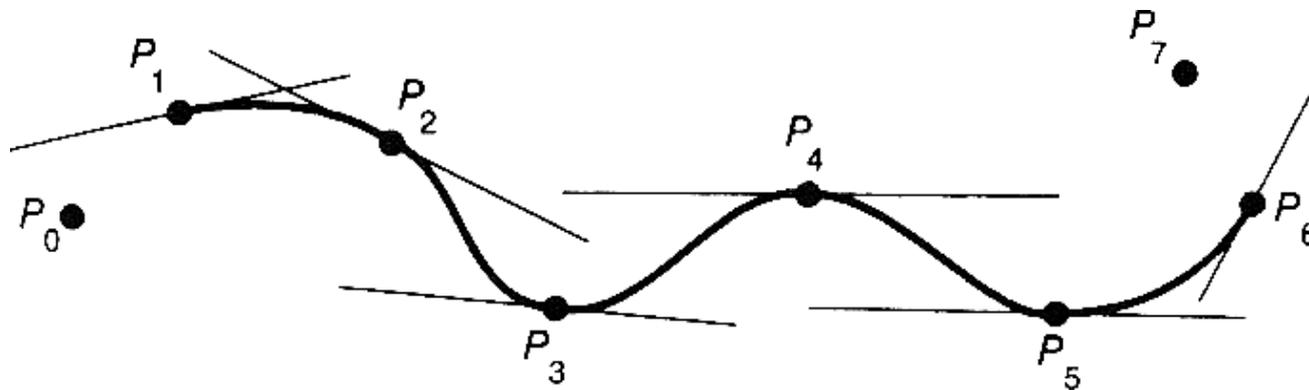


$$C = H \cdot T_{CH} = \frac{1}{2} \begin{pmatrix} -1 & 3 & -3 & 1 \\ 2 & -5 & 4 & -1 \\ -1 & 0 & 1 & 0 \\ 0 & 2 & 0 & 0 \end{pmatrix}$$



Catmull-Rom-Spline

- Piecewise polynomial curve
- Four control points per segment
- For n control points we obtain $(n - 3)$ polynomial segments



Application

- Smooth interpolation of a given sequence of points
- Key frame animation, camera movement, *etc.*
- Only G^1 -continuity
- Control points should be equidistant in time



Problem

- Often only the control points are given
- How to obtain a suitable parameterization t_i ?

Example: *Chord-Length Parameterization*

$$t_0 = 0$$

$$t_i = \sum_{j=1}^i \text{dist}(P_j - P_{j-1})$$

- Arbitrary up to a constant factor

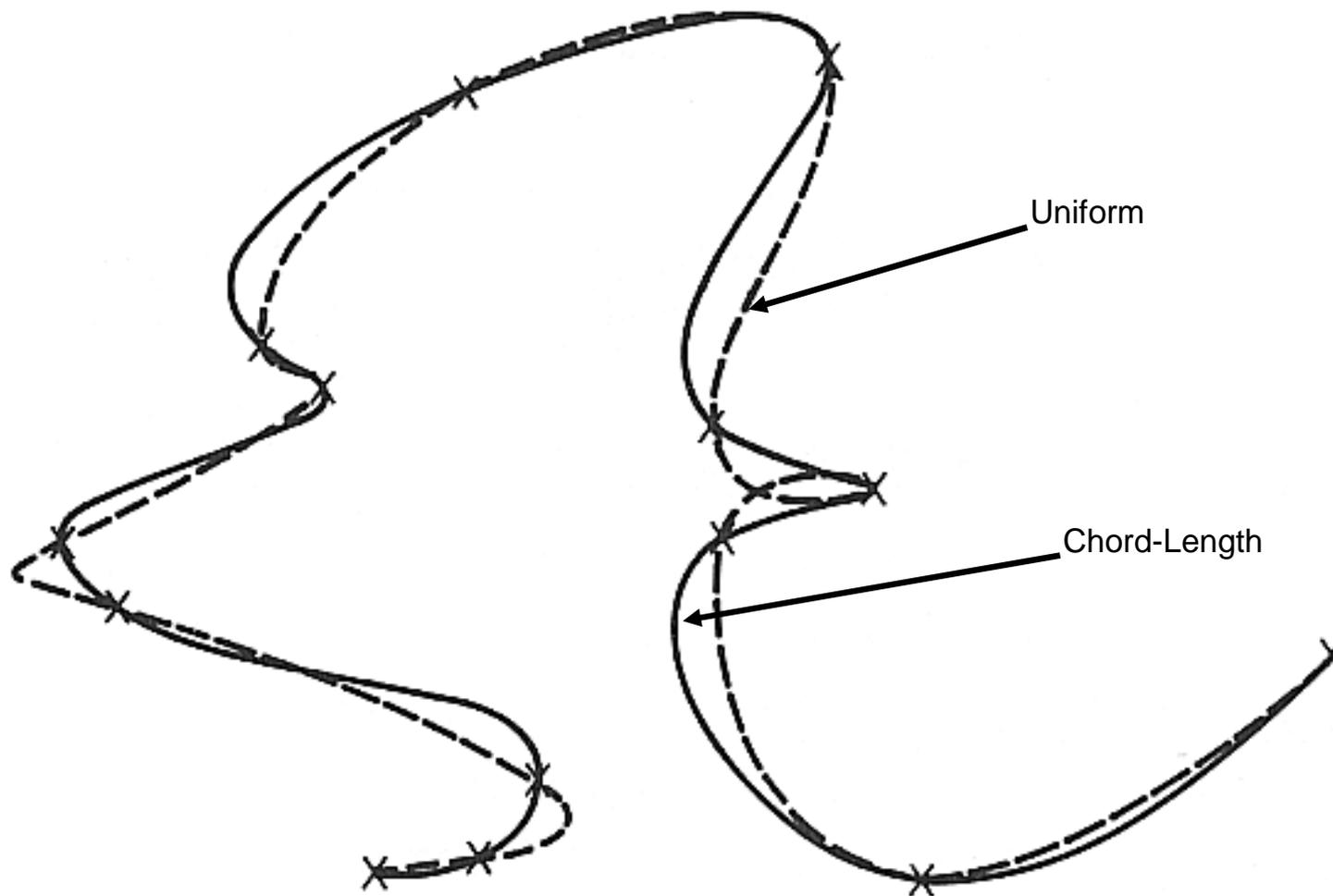
Warning

- Distances are not affine invariant !
- Shape of curves changes under transformations !!



Chord-Length versus uniform Parameterization

- Analog: Think $P(t)$ as a moving object with mass that may overshoot





Goal

- Spline curve with local control and high continuity

Given

- Degree: n
- Control points: p_0, \dots, p_m (Control polygon, $m \geq n + 1$)
- Knots: t_0, \dots, t_{m+n+1} (Knot vector, weakly monotonic)
- The knot vector defines the parametric locations where segments join

B-Spline Curve

$$P(t) = \sum_{i=0}^m N_i^n(t) p_i$$

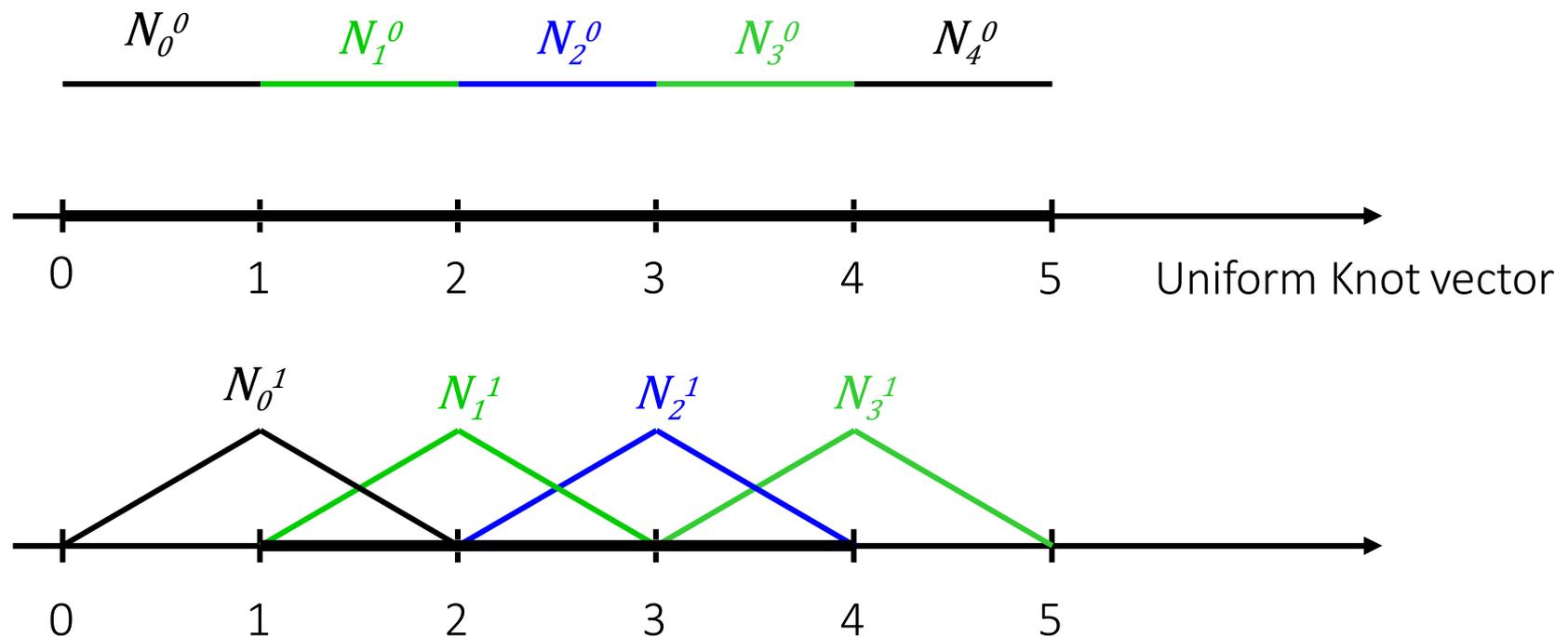
- Continuity:
 - C^{n-1} at simple knots
 - C^{n-k} at knot with multiplicity k



Recursive Definition

$$N_i^0(t) = \begin{cases} 1 & \text{if } t_i < t < t_{i+1} \\ 0 & \text{otherwise} \end{cases}$$

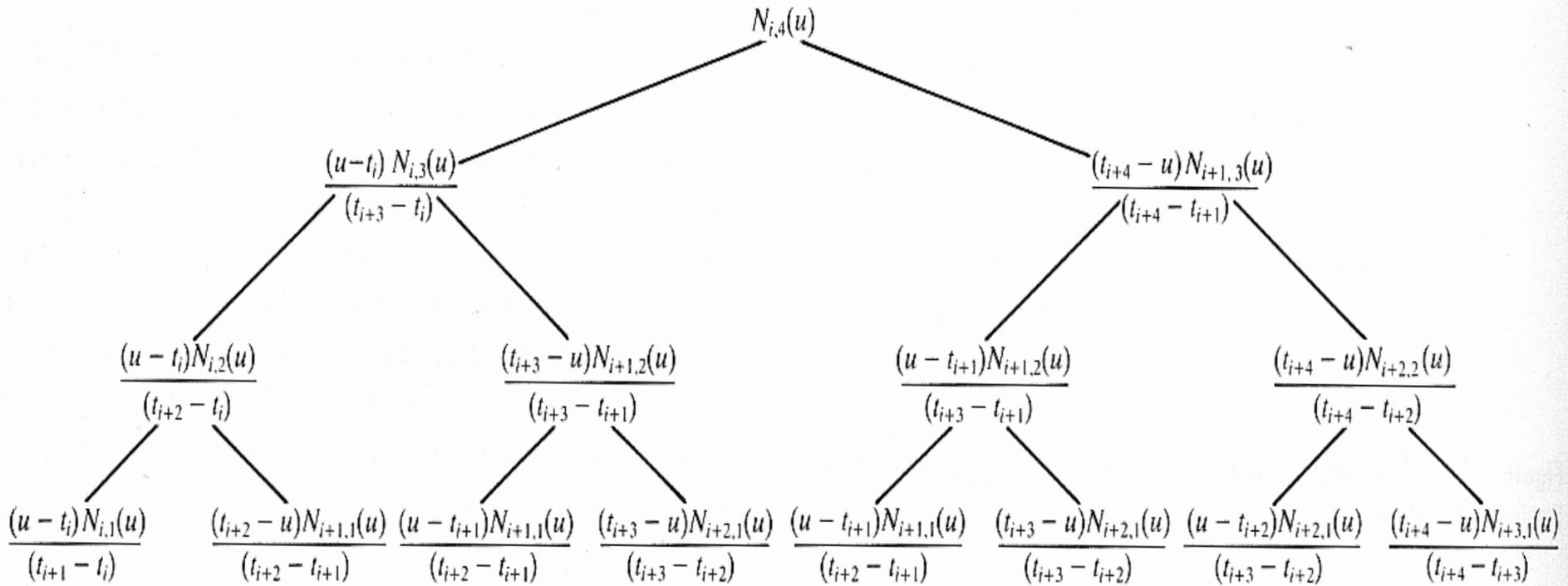
$$N_i^n(t) = \frac{t - t_i}{t_{i+n} - t_i} N_i^{n-1}(t) + \frac{t_{i+1} - t}{t_{i+1} - t_{i+2}} N_{i+1}^{n-1}(t)$$





Recursive Definition

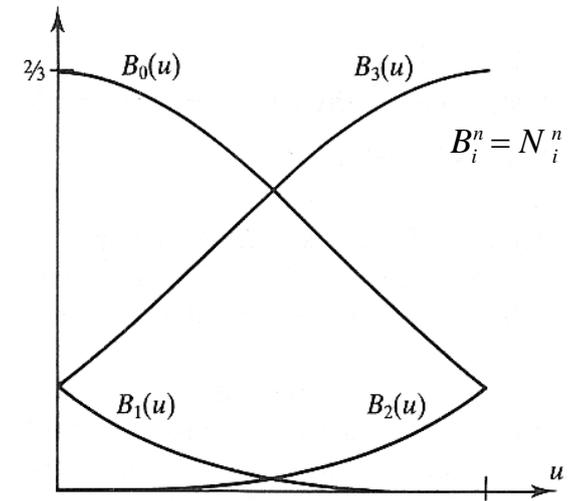
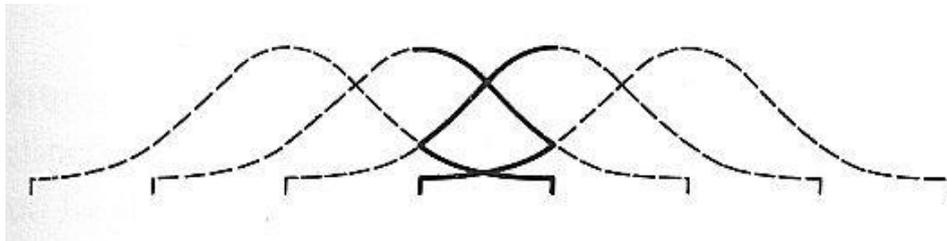
- Degree increases in every step
- Support increases by one knot interval





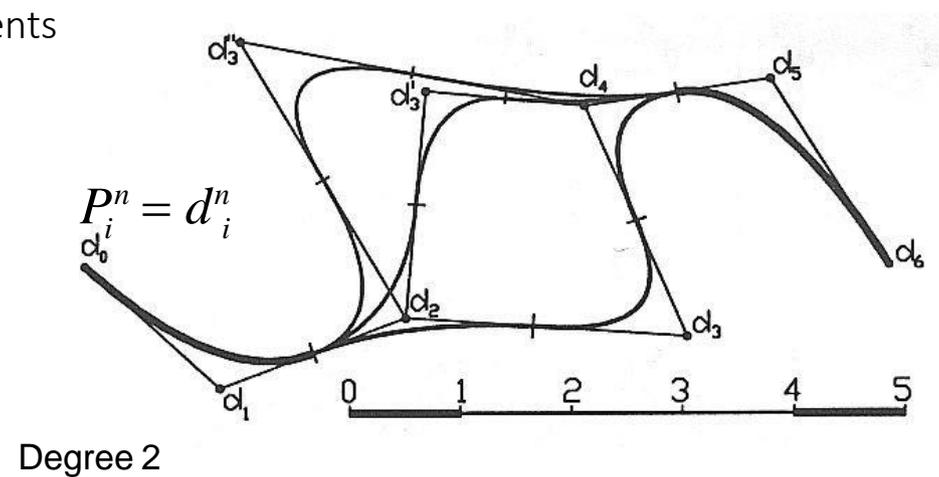
Uniform Knot Vector

- All knots at integer locations
 - UBS: Uniform B-Spline
- Example: cubic B-Splines



Local Support = Localized Changes

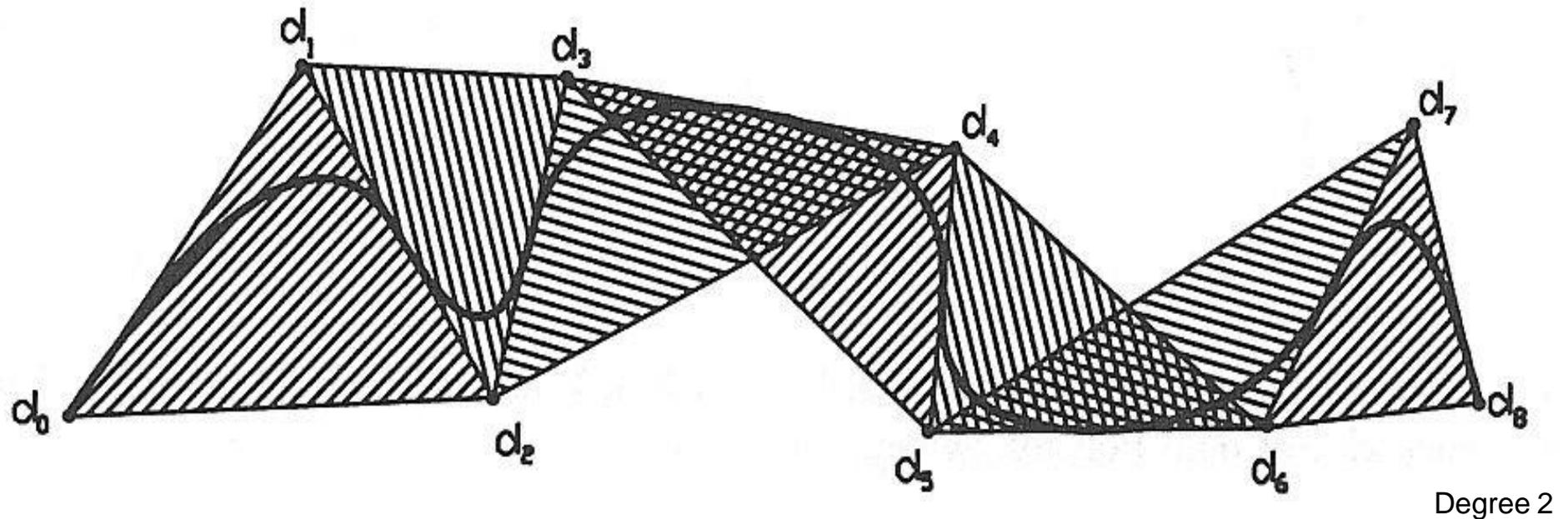
- Basis functions affect only $(n + 1)$ Spline segments
- Changes are localized





Convex Hull Property

- Spline segment lies in convex Hull of $(n+1)$ control points

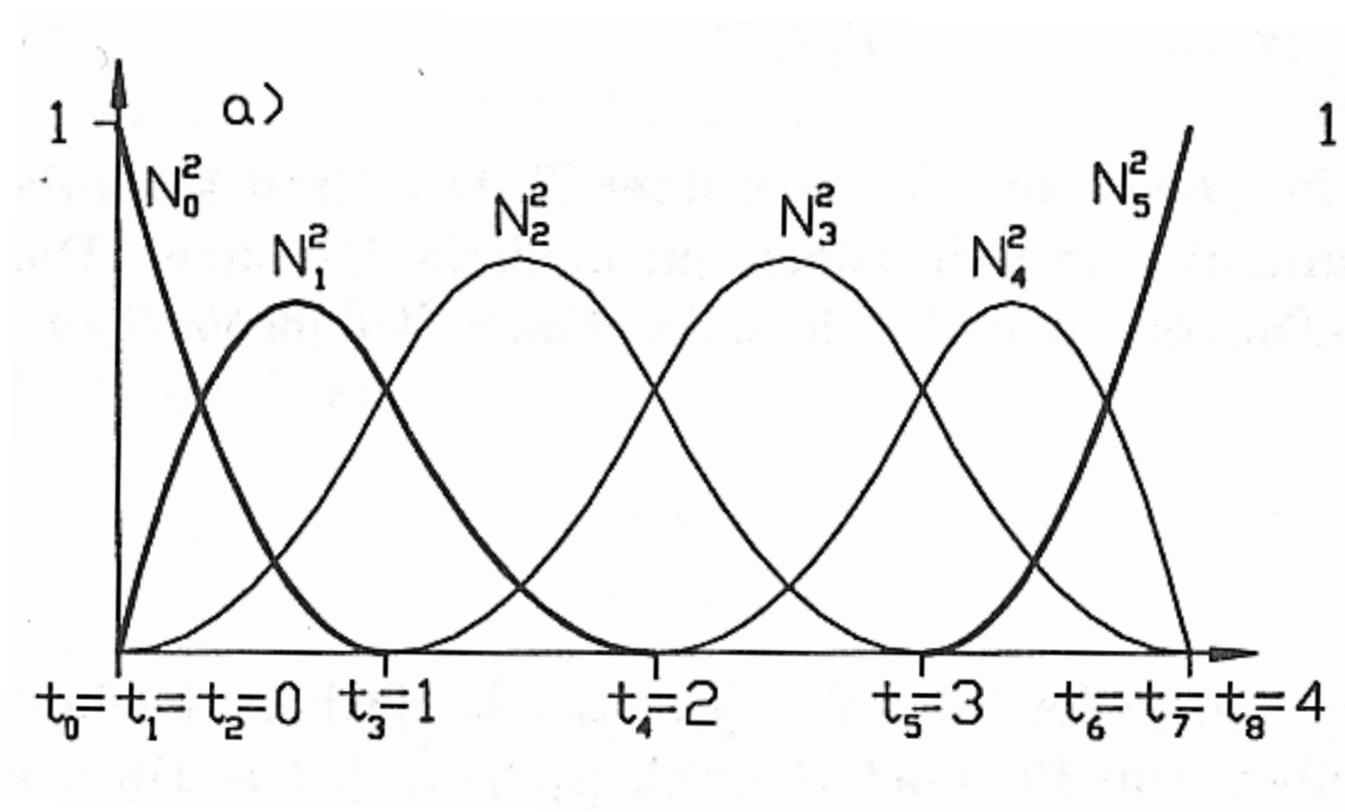


- $(n + 1)$ control points lie on a straight line \Rightarrow curve touches this line
- n control points coincide \Rightarrow curve interpolates this point and is tangential to the control polygon



Basis Functions on an Interval

- Knots at beginning and end with multiplicity
 - NUBS: Non-uniform B-Splines
- Interpolation of end points and tangents there
- Conversion to Bézier segments via **knot insertion**





Recursive Definition of Control Points

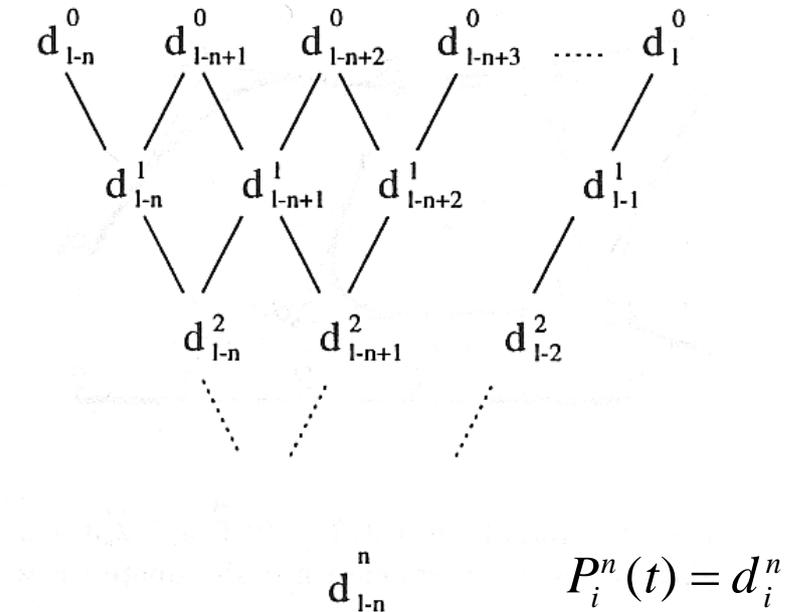
- Evaluation at t : $t_i < t < t_{i+1}: i \in \{l-n, \dots, l\}$
- Due to local support only affected by $(n+1)$ control points

$$P_i^r(t) = \left(1 - \frac{t - t_{i+r}}{t_{i+n+1} - t_{i+r}}\right) P_i^{r-1}(t) + \frac{t - t_{i+r}}{t_{i+n+1} - t_{i+r}} P_{i+1}^{r-1}(t)$$

$$P_i^0(t) = P_i$$

Properties

- Affine invariance
- Stable numerical evaluation
 - All coefficients > 0





Algorithm similar to deBoor

- Given a new knot t
 - $t_i < t < t_{i+1} : i \in \{l - n, \dots, l\}$
- $T^* = T \cup \{t\}$
- New representation of the same curve over T^*

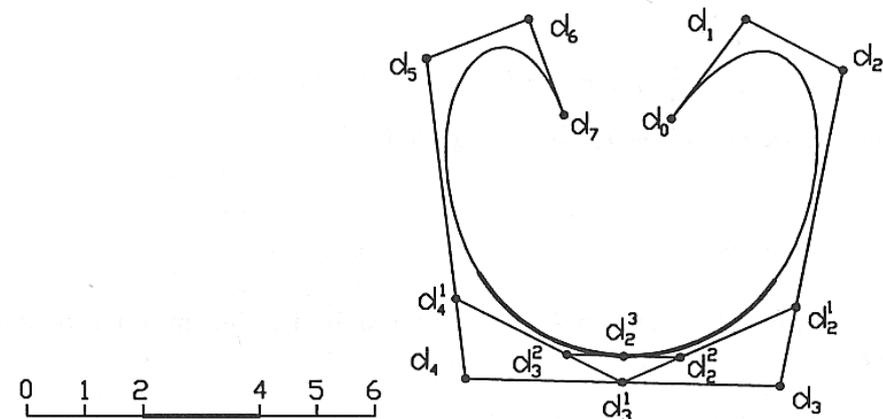
$$P^*(t) = \sum_{i=0}^{m+1} N_i^n(t) P_i^*$$

$$P_i^* = (1 - a_i) P_{i-1} + a_i P_i$$

$$a_i = \begin{cases} 1 & i \leq l - n \\ \frac{t - t_i}{t_{i+n} - t_i} & l - n + 1 \leq i \leq l \\ 0 & i \geq l + 1 \end{cases}$$

Applications

- Refinement of curve, display



Consecutive insertion of three knots at $t = 3$ into a cubic B-Spline.

First and last knot have multiplicity n

$T = (0,0,0,0,1,2,4,5,6,6,6,6), l = 5$





B-Spline to Bézier Representation

- Remember:
 - Curve interpolates point and is tangential at knots of multiplicity n
- In more detail: If two consecutive knots have multiplicity n
 - The corresponding spline segment is in Bézier form
 - The $(n + 1)$ corresponding control polygon form the Bézier control points

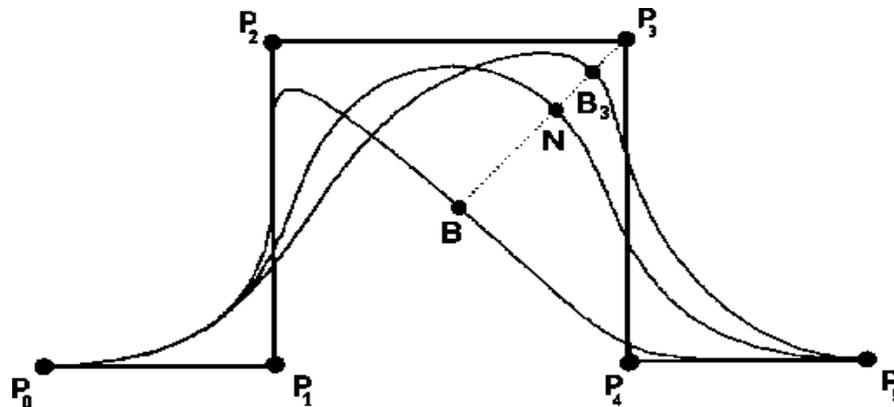


Non-Uniform Rational B-Splines

- Homogeneous control points: now with weight w_i
 - $P'_i = (w_i x_i, w_i y_i, w_i z_i, w_i) = w_i P_i$

$$P'(t) = \sum_{i=0}^m N_i^n(t) P'_i$$

$$P = \frac{\sum_{i=0}^m N_i^n(t) P_i w_i}{\sum_{i=0}^m N_i^n(t) w_i} = \sum_{i=0}^m R_i^n(t) P_i w_i, \quad \text{with } R_i^n(t) = \frac{N_i^n(t) w_i}{\sum_{i=0}^m N_i^n(t) w_i}$$





Properties

- Piecewise rational functions
- Weights
 - High (relative) weight attract curve towards the point
 - Low weights repel curve from a point
 - Negative weights should be avoided (may introduce singularity)
- Invariant under projective transformations
- Variation-Diminishing-Property (in functional setting)
 - Curve cuts a straight line no more than the control polygon does

Examples: Cubic B-Splines

